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# Robust Global Exponential Stabilization on the $n$ -Dimensional Sphere with Applications to Trajectory Tracking for Quadrotors<sup>★</sup>

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## Abstract

In this paper, we design a hybrid controller that globally exponentially stabilizes a system evolving on the  $n$ -dimensional sphere, denoted by  $\mathbb{S}^n$ . This hybrid controller is induced by a “synergistic” collection of potential functions on  $\mathbb{S}^n$ . We propose a particular construction of this class of functions that generates flows along geodesics of the sphere, providing convergence to the desired reference with minimal path length. We show that the proposed strategy is suitable to the exponential stabilization of a quadrotor vehicle.

*Key words:* Hybrid control systems, Synergistic potential functions, Lyapunov-based control.

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## 1 Introduction

### 1.1 Motivation and Problem Statement

In this paper, we design a hybrid controller for global exponential stabilization of a setpoint for a system evolving on the  $n$ -dimensional sphere, given by  $\mathbb{S}^n := \{x \in \mathbb{R}^{n+1} : x^\top x = 1\}$ . The dynamics of this system can be

described by

$$\dot{x} = \Pi(x)\omega \quad x \in \mathbb{S}^n, \quad (1)$$

where  $\omega \in \mathbb{R}^{n+1}$  is the input and  $\Pi(x) = I_n - xx^\top$  projects  $\omega$  onto the tangent space to  $\mathbb{S}^n$  at  $x$ , given by  $\mathbb{T}_x \mathbb{S}^n := \{z \in \mathbb{R}^{n+1} : z^\top x = 0\}$ . Even though there exist controllers that globally asymptotically stabilize a setpoint on the  $n$ -dimensional sphere (c.f. Mayhew and Teel [2013a]) and others that globally exponentially stabilize a setpoint on the special orthogonal group of order 3 (c.f. Lee [2015] and Berkane et al. [2017]), to the best of our knowledge, the problem of global exponential stabilization of a setpoint in  $\mathbb{S}^n$  has not been addressed before, despite the fact that it has many meaningful applications, such as visual servoing (c.f. Triantafyllou et al. [2018]); control of robotic manipulators (c.f. Chaturvedi and McClamroch [2009]); exoskeleton tracking (c.f. Brahmi et al. [2019]); multi-agent synchronization (c.f. Markdahl et al. [2018]); formation control (c.f. Zhao and Zelazo [2016]); rigid-body stabilization (c.f. Chaturvedi et al. [2011]) and trajectory tracking for multi-rotor aerial vehicles (c.f. Mahony et al. [2012]), which we also explore in this paper. To see how the proposed controller for global exponential stabilization on the  $n$ -dimensional sphere applies to trajectory tracking for a multi-rotor aerial vehicle, consider the following: the position dynamics of

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a multi-rotor vehicle can be described by

$$\dot{p} = v \quad \dot{v} = xu + g$$

where  $g$  is the acceleration of gravity,  $p \in \mathbb{R}^3$  and  $v \in \mathbb{R}^3$  denote the position and the velocity of the vehicle with respect to the inertial reference frame,  $u$  denotes the magnitude of the thrust and  $x \in \mathbb{S}^2$  denotes the direction of the thrust (c.f. Hamel et al. [2002]). Given a reference trajectory with acceleration  $\ddot{p}_d$  and a control law  $w : \mathbb{R}^3 \rightarrow \mathbb{R}^3$  that exponentially stabilizes the double integrator, if the controller for  $x$  exponentially stabilizes the commanded thrust direction, given by  $\rho(\tilde{p}, \tilde{v}, \ddot{p}_d) := \frac{w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d}{|w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d|}$  for each  $(\tilde{p}, \tilde{v}, \ddot{p}_d) \in \{(\tilde{p}, \tilde{v}, \ddot{p}_d) \in \mathbb{R}^9 : w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d \neq 0\}$  where  $\tilde{p}$  and  $\tilde{v}$  denote the position and velocity tracking errors, respectively, then exponential tracking is attained. This, however, cannot be achieved through continuous feedback because it is not possible to globally asymptotically stabilize a given setpoint on a compact manifold by means of continuous feedback (c.f. Bhat and Bernstein [2000]). Moreover, if a dynamical system cannot be globally asymptotically stabilized through continuous feedback, it cannot be robustly stabilized by discontinuous feedback either, as shown in Mayhew and Teel [2011]. Overcoming these limitations is particularly important for multi-rotor aerial vehicles due to their popularity and the wide range of applications in which they are used, such as surveillance, tracking, search and rescue, infrastructure inspection, agriculture and disaster mitigation (see e.g. Fink et al. [2011], Mellinger et al. [2012], Jiang and Kumar [2013], Augugliaro et al. [2013], Lupashin et al. [2014], Floreano and Wood [2015] and Liang et al. [2018]). In this paper, we provide a hybrid controller that not only provides global exponential stability, but also confers a quantifiable robustness margin to perturbations.

Recent developments on hybrid control theory overcame the topological obstructions to global stabilization on compact manifolds with the introduction of synergistic potential functions in Mayhew and Teel [2010]. A potential function on a compact manifold is a continuously differentiable function that is positive definite relative to a given point, thus it induces a gradient vector field which asymptotically stabilizes the given reference from every initial condition except a set of measure zero. To see this, consider the closed-loop system resulting from the interconnection of the gradient-based feedback law of a potential function  $h_r$  that is positive definite relative to  $r \in \mathbb{S}^n$  and (1), given by:

$$\dot{x} = -\Pi(x)\nabla h_r(x). \quad (2)$$

The equilibrium points of (2) are the critical points of  $h_r(x)$ , denoted by

$$\text{crit } h_r := \{x \in \mathbb{S}^n : \Pi(x)\nabla h_r(x) = 0\}, \quad (3)$$

which correspond to the set of points  $x$  where  $\nabla h_r(x)$  is orthogonal to  $T_x \mathbb{S}^n$ . Since the set (3) includes the maxi-

mum and the minimum of  $h_r$  on  $\mathbb{S}^n$ , it follows that  $r \in \mathbb{S}^n$  is not globally asymptotically stable for (2). On the other hand, synergistic potential functions are collections of potential functions that enable a controller to achieve robust global asymptotic stabilization of the given setpoint, because, at the undesired equilibria, there exists another function in the collection with a lower value that we can switch to.

The concept of synergistic potential functions has been introduced to address the problem of stabilizing a three-dimensional pendulum in Mayhew and Teel [2010] and later used in full attitude stabilization in Mayhew and Teel [2013b], Lee [2015]; attitude synchronization Mayhew et al. [2012]; partial attitude stabilization (c.f. Mayhew and Teel [2013a]), and stabilization by hybrid backstepping (c.f. Mayhew et al. [2011b]). More recently, the interest in this control technique has spawned the design of new synergistic potential functions on  $\text{SO}(3)$ , such as the ones by Berkane and Tayebi [2015], Berkane and Tayebi [2017] and Berkane et al. [2017]. It was shown by the authors in Casau et al. [2015b] that synergistic potential functions can be used for global asymptotic stabilization of a reference trajectory for a multi-rotor aerial vehicle, but the extent to which exponential stabilization is possible was not addressed.

## 1.2 Contributions

The contributions in this paper are as follows. *Extension to the concept of synergistic potential functions:* in Section 3, we show that the existence of a centrally synergistic potential function  $V : \mathbb{S}^n \times \mathcal{Q} \rightarrow \mathbb{R}_{\geq 0}$ , where  $\mathcal{Q}$  is a compact set, induces a gradient-based control law that renders

$$\mathcal{A} := \{(x, y) \in \mathbb{S}^n \times \mathcal{Q} : x = r\} \quad (4)$$

globally asymptotically stable for the closed-loop system. This nomenclature is inherited from Mayhew and Teel [2013a], where synergistic potential functions satisfying  $V(r, y) = 0$  for all  $y \in \mathcal{Q}$  are said to be central because they share a common minimum at  $r$ . In this paper, we extend the previous notion of centrally synergistic potential functions, because we consider that  $\mathcal{Q}$  is compact rather than finite, which adds flexibility to the design of synergistic potential functions. The proposed controller is significantly different from the one in Mayhew and Teel [2013b] because the focus is on global exponential stabilization of  $\mathbb{S}^n$  rather than  $\text{SO}(3)$  and it further expands the work in Mayhew and Teel [2010] from global asymptotic stabilization on  $\mathbb{S}^2$  to global exponential stabilization on  $\mathbb{S}^n$ . Interestingly, the controller that we propose may be used for global stabilization on  $\mathbb{S}^3$  which is the universal cover of  $\text{SO}(3)$ . However, the resulting controller would be more complex than that of Mayhew et al. [2011a] if used for rigid-body stabilization, because it would not take advantage of the fact that  $\mathbb{S}^3$  is a double cover of  $\text{SO}(3)$ . *Global exponential stability of a setpoint on  $\mathbb{S}^n$ :* in Theorem 2,

we show that, if  $V$  is bounded from above and below by a polynomial function of the distance to  $\mathcal{A}$  and if  $V$  converges exponentially fast to 0, then  $\mathcal{A}$  is globally exponentially stable for the closed-loop system, in the sense that, for all initial conditions, the state of the system converges exponentially to  $\mathcal{A}$ . This is different from the controllers proposed in Lee [2015] and Berkane et al. [2017], because these address the problem of global exponential stabilization on  $\text{SO}(3)$  rather than  $\mathbb{S}^n$ . *Optimal switching*: in Section 3.1, we construct a synergistic potential function on the  $n$ -dimensional sphere that meets the requirements for exponential stability and has an optimal switching law, in the sense that it guarantees that solutions to the closed-loop system follow geodesics whenever a jump of the hybrid controller is triggered. *Robustness to perturbations*: the proposed hybrid controller satisfies the so-called hybrid basic conditions, therefore it is endowed with nominal robustness properties that are outlined in Goebel et al. [2012]. In addition, the switching rule introduces a hysteresis gap that prevents chattering. *Saturated thrust feedback for exponential tracking of a reference trajectory for a multi-rotor aerial vehicle*: in Section 4, we employ the hybrid controller for global exponential stabilization on  $\mathbb{S}^n$  in trajectory tracking for a multi-rotor aerial vehicle. We show that, for each compact set of initial position and velocity tracking errors and for all initial orientations, the reference trajectory is exponentially stable. This is particularly difficult, because in addition to the topological constraints to global attitude stabilization, multirotor aerial vehicles are subject to underactuation constraints that prevent stabilization of the vehicle when the commanded thrust is zero (c.f. Lizárraga [2004]). This issue has been widely acknowledged but mostly overlooked due to its singular nature. For example, in Lee et al. [2013] this flight condition is assumed to not occur and in Hua et al. [2009] the controller is turned off when the commanded thrust approaches zero. Similarly to the work of Hua et al. [2015], we assume that the reference trajectory does not lead to a situation where the commanded thrust is zero, but, unlike the aforementioned approach, we explicitly build this limitation into the control design procedure so to achieve semi-global exponential stability with respect to the position and velocity errors. A video of experimental runs using this controller can be found in Casau et al. [2017a].

The paper is organized as follows: in Section 2, we present the notation and the framework of hybrid dynamical systems that is used in this paper. In Section 3, we develop the notions of synergistic potential functions on the  $n$ -dimensional sphere. In Section 4, we apply the given controllers to the tracking of a reference trajectory for a vectored-thrust vehicle, and in Section 5, we present the conclusions of this work. In Casau et al. [2015a] we reported the results that are presented in Section 3 without the proofs. In Casau et al. [2016], we presented some preliminary results on work reported in Section 4 without the constructive controller synthesis that is presented in this paper for the quadrotor application.

## 2 Preliminaries

The symbol  $\mathbb{R}$  denotes the set of real numbers,  $\mathbb{N}$  denotes the set of natural numbers and zero,  $\mathbb{R}_{\geq 0} := \{x \in \mathbb{R} : x \geq 0\}$ ,  $\mathbb{R}^n$  denotes the  $n$ -dimensional Euclidean space equipped with the norm  $|x| := \sqrt{\langle x, x \rangle}$  for each  $x \in \mathbb{R}^n$ , where  $\langle u, v \rangle := u^\top v$  for each  $u, v \in \mathbb{R}^n$ . The canonical basis for  $\mathbb{R}^n$  is denoted by  $\{e_i\}_{1 \leq i \leq n} \subset \mathbb{R}^n$  and  $c+r\mathbb{B} := \{x \in \mathbb{R}^n : |x - c| \leq r\}$ . If a real symmetric matrix  $A \in \mathbb{R}^{n \times n}$  is positive (negative) definite, we write  $A \in \mathbb{S}_{>0}^n$  ( $A \in \mathbb{S}_{<0}^n$ ) or  $A \succ 0$  ( $A \prec 0$ ) if the dimensions can be inferred from context. If a real symmetric matrix  $A \in \mathbb{R}^{n \times n}$  is positive (negative) semidefinite, we write  $A \in \mathbb{S}_{\geq 0}^n$  ( $A \in \mathbb{S}_{\leq 0}^n$ ) or  $A \succeq 0$  ( $A \preceq 0$ ) if the dimensions can be inferred from context. Given  $A \in \mathbb{R}^{m \times n}$ ,  $\sigma_{\max}(A)$  denotes the maximum singular value of  $A$ . The gradient of a continuously differentiable function  $V : \mathbb{R}^n \rightarrow \mathbb{R}$  is given by  $\nabla V(x) := \left[ \frac{\partial V}{\partial x_1}(x) \dots \frac{\partial V}{\partial x_n}(x) \right]$  for each  $x = (x_1, \dots, x_n) \in \mathbb{R}^n$ . The derivative of a differentiable matrix function with matrix arguments  $F : \mathbb{R}^{m \times n} \rightarrow \mathbb{R}^{k \times \ell}$  is given by  $\mathcal{D}_X(F(X)) := \partial \text{vec}(F(X)) / \partial \text{vec}(X)^\top$  for each  $X \in \mathbb{R}^{m \times n}$ . The domain of a set-valued mapping  $M : \mathbb{R}^n \rightrightarrows \mathbb{R}^m$  is given by  $\text{dom } M := \{x \in \mathbb{R}^n : M(x) \neq \emptyset\}$ . The range of  $M$  is the set  $\text{rge } M := \{y \in \mathbb{R}^m : \exists x \in \mathbb{R}^n \text{ such that } y \in M(x)\}$ . A hybrid system  $\mathcal{H}$  defined on  $\mathbb{R}^n$  can be represented by

$$\mathcal{H} : \begin{cases} \dot{x} \in F(x) & x \in C \\ x^+ \in G(x) & x \in D \end{cases} \quad (5)$$

where  $C \subset \mathbb{R}^n$  is the flow set,  $F : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  is the flow map,  $D \subset \mathbb{R}^n$  and  $G : \mathbb{R}^n \rightrightarrows \mathbb{R}^n$  with  $D \subset \text{dom } G$  is the jump map, as defined in Goebel et al. [2012]. The maps  $F$  and  $G$  are set-valued maps satisfying  $C \subset \text{dom } F$  and  $D \subset \text{dom } G$ , respectively. Loosely speaking, solutions to hybrid systems are hybrid arcs that are compatible with the data of the hybrid system, i.e., functions  $(t, j) \mapsto \phi(t, j)$  defined on a hybrid time domain  $E \subset \mathbb{R}_{\geq 0} \times \mathbb{N}_0$  that satisfy  $\dot{\phi}(t, j) \in F(\phi(t, j))$  for almost all  $t \geq 0$  and  $\phi(t, j) \in D$ ,  $\phi(t, j+1) \in G(\phi(t, j))$  for every  $\phi(t, j)$  and  $\phi(t, j+1)$  belonging to  $\text{dom } \phi$ . We say that a solution  $\phi$  to (5) is maximal if it cannot be extended, it is complete if its domain is unbounded, it is discrete if  $\text{dom } \phi \subset \{0\} \times \mathbb{N}$  and it is Zeno if it is complete and  $\sup_t \text{dom } \phi < +\infty$  (for a more rigorous description of solutions to hybrid systems see Goebel et al. [2012]). Under the assumption of completeness of maximal solution to  $\mathcal{H}$ , a compact set  $\mathcal{A}$  is said to be: stable for  $\mathcal{H}$ , if for each  $\epsilon > 0$  there exists  $\delta > 0$  such that for each solution  $\phi$  to  $\mathcal{H}$  with  $|\phi(0, 0)|_{\mathcal{A}} \leq \delta$  satisfies  $|\phi(t, j)|_{\mathcal{A}} \leq \epsilon$  for each  $(t, j) \in \text{dom } \phi$ ; attractive for  $\mathcal{H}$  if  $\lim_{t+j \rightarrow \infty} |\phi(t, j)|_{\mathcal{A}} = 0$ , where  $|x|_{\mathcal{A}} := \min_{y \in \mathcal{A}} |x - y|$  and  $|x| := \sqrt{\langle x, x \rangle}$  for each  $x \in \mathbb{R}^n$ . A set  $\mathcal{A}$  is globally asymptotically stable for  $\mathcal{H}$  if it is both globally attractive and globally stable for  $\mathcal{H}$ . We say that a compact set  $\mathcal{A} \subset \mathbb{R}^n$  is exponentially stable in the  $t$ -direction from  $U$  if there exists  $k, \lambda > 0$  such that  $\sup_t \text{dom } \phi = \infty$  and  $|\phi(t, j)|_{\mathcal{A}} \leq k \exp(-\lambda t) |\phi(0, 0)|_{\mathcal{A}}$  for each maximal solution  $\phi$  to the hybrid system with

$\phi(0, 0) \in U$  and  $(t, j) \in \text{dom } \phi$ .

### 3 Synergistic Potential Functions on $\mathbb{S}^n$

In this section, we design a hybrid controller that *globally* exponentially stabilizes a given reference  $r \in \mathbb{S}^n := \{x \in \mathbb{R}^{n+1} : x^\top x = 1\}$  for the system (1) using the notion of centrally synergistic potential functions given next.

**Definition 1** A function  $h_r : \mathbb{S}^n \rightarrow \mathbb{R}$  is said to be a *potential function on  $\mathbb{S}^n$  relative to  $r$*  if it is continuously differentiable and positive definite relative to  $r \in \mathbb{S}^n$ , i.e.,  $h_r(x) \geq 0$  for each  $x \in \mathbb{S}^n$  and  $h_r(x) = 0$  if and only if  $x = r$ . We denote the collection of potential functions on  $\mathbb{S}^n$  relative to  $r$  by  $\mathcal{P}_r$ .  $\square$

**Definition 2** Given a compact set  $\mathcal{Q}$ , a function  $V : \mathbb{S}^n \times \mathcal{Q} \rightarrow \mathbb{R}$  is said to be a *centrally synergistic potential function on  $\mathbb{S}^n$  relative to  $r$*  if the following holds: (1) For each  $y \in \mathcal{Q}$ ,  $V^y$  is a potential function on  $\mathbb{S}^n$  relative to  $r$ , where  $V^y(x) := V(x, y)$  for each  $x \in \mathbb{S}^n$ ; (2) There exists  $\delta > 0$  such that

$$\mu(x, y) := V(x, y) - \nu(x) > \delta \quad (6)$$

for each  $(x, y) \in \mathcal{E}(V)$ , where

$$\mathcal{E}(V) := \{(x, y) \in \mathbb{S}^n \times \mathcal{Q} : \Pi(x) \nabla V^y(x) = 0, x \neq r\} \quad (7)$$

and

$$\nu(x) := \min_{\bar{y} \in \mathcal{Q}} V(x, \bar{y}). \quad (8)$$

We denote the collection of centrally synergistic potential functions on  $\mathbb{S}^n$  relative to  $r$  by  $\mathcal{Q}_r$ .  $\square$

Given a centrally synergistic potential function  $V$ , the function  $(x, y) \rightarrow \mu(x, y)$  is referred to as the *synergy gap* of  $V$  at  $(x, y)$ . The synergy gap measures the difference between the current value of the function and all other potential functions in the collection. If (6) holds, we say that  $V$  has *synergy gap exceeding  $\delta$* . Given  $V \in \mathcal{Q}_r$ , we define the hybrid controller with output  $\omega \in \mathbb{R}^{n+1}$ , state  $y \in \mathcal{Q}$  and input  $x \in \mathbb{S}^n$  as follows:

$$\left. \begin{aligned} \omega &= -\nabla V^y(x) \\ \dot{y} &= 0 \end{aligned} \right\} (x, y) \in C := \{(x, y) \in \mathcal{X} : \mu(x, y) \leq \delta\} \quad (9a)$$

$$y^+ \in \varrho(x) \quad (x, y) \in D := \{(x, y) \in \mathcal{X} : \mu(x, y) \geq \delta\}, \quad (9b)$$

where  $\mathcal{X} := \mathbb{S}^n \times \mathcal{Q}$  and

$$\varrho(x) := \arg \min_{\bar{y} \in \mathcal{Q}} V(x, \bar{y}). \quad (10)$$

The interconnection between (1) and (9) is represented by the closed-loop hybrid system  $\mathcal{H} := (C, F, D, G)$ ,

given by

$$\begin{aligned} \begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} &= F(x, y) = \begin{pmatrix} -\Pi(x) \nabla V^y(x) \\ 0 \end{pmatrix} & (x, y) \in C \\ \begin{pmatrix} x^+ \\ y^+ \end{pmatrix} &\in G(x, y) = \begin{pmatrix} x \\ \varrho(x) \end{pmatrix} & (x, y) \in D. \end{aligned} \quad (11)$$

The parameter  $\delta$  that is used in the hybrid controller (9) is central to the construction of synergistic potential functions and it defines a hysteresis gap which prevents chattering. This construction is similar across many earlier works on synergistic hybrid feedback (see e.g., Mayhew and Teel [2013a], Mayhew and Teel [2013b]) and it guarantees global asymptotic stability of  $x = r$  for the hybrid system (11), as shown next.

**Theorem 1** Given  $r \in \mathbb{S}^n$  and a compact set  $\mathcal{Q}$ , if there exists  $\delta > 0$  such that  $V \in \mathcal{Q}_r$  has synergy gap exceeding  $\delta$ , then the set  $\mathcal{A}$  in (4) is globally asymptotically stable for the hybrid system  $\mathcal{H}$  in (11).

**PROOF.** It follows from Proposition A.1 that [Goebel et al., 2012, Assumption 6.5] is satisfied. From computations similar to Mayhew and Teel [2013a], we conclude that the growth of  $V$  along solutions to (11) is bounded by  $u_C, u_D$ , where

$$u_C(x, y) := \begin{cases} -|\Pi(x) \nabla V^y(x)|^2 & \text{if } (x, y) \in C \\ -\infty & \text{otherwise} \end{cases}, \quad (12a)$$

$$u_D(x, y) := \begin{cases} -\delta & \text{if } (x, y) \in D \\ -\infty & \text{otherwise} \end{cases} \quad (12b)$$

for each  $(x, y) \in \mathbb{S}^n \times \mathcal{Q}$ . It follows from the assumption that  $V$  is a synergistic potential function relative to  $r$  that  $\mu(x, y) > \delta$  for each  $(x, y) \in \mathcal{E}(V)$ , hence  $u_C(x, y) < 0$  for each  $(x, y) \in \mathbb{S}^n \times \mathcal{Q} \setminus \mathcal{A}$ . Since  $u_D(x, y) < 0$  for all  $(x, y) \in \mathbb{S}^n \times \mathcal{Q}$ , it follows from [Goebel et al., 2012, Corollary 8.9] that  $\mathcal{A}$  is globally pre-asymptotically stable for (11). Since  $G(D) \subset C \cup D = \mathbb{S}^n \times \mathcal{Q}$ ,  $\mathbb{S}^n \times \mathcal{Q}$  is compact and, for each  $(x, y) \in C \setminus D$ ,  $F(x, y)$  belongs to the tangent cone to  $C$  at  $(x, y)$ , it follows from [Goebel et al., 2012, Proposition 6.10] that each maximal solution to (11) is complete.  $\square$

We are also able to show that, under some additional conditions, the set (4) is globally exponentially stable in the  $t$ -direction for the hybrid system (11).

**Theorem 2** Given  $r \in \mathbb{S}^n$  and a compact set  $\mathcal{Q}$ , if there exists  $V \in \mathcal{Q}_r$  with synergy gap exceeding  $\delta > 0$  satisfying the following conditions

$$\underline{\alpha} |x - r|^p \leq V(x, y) \leq \bar{\alpha} |x - r|^p \quad \forall (x, y) \in C \cup D, \quad (13a)$$

$$\langle \nabla V(x, y), F(x, y) \rangle \leq -\lambda V(x, y) \quad \forall (x, y) \in C, \quad (13b)$$



for some  $p, \underline{\alpha}, \bar{\alpha}, \lambda > 0$ , then the set (4) is globally exponentially stable in the  $t$ -direction for the hybrid system (11).

**PROOF.** It follows from the proof of Theorem 1 that (11) satisfies [Goebel et al., 2012, Assumption 6.5], each of its maximal solutions is complete and  $V$  is nonincreasing during jumps. Since  $\mu(g) = 0$  for each  $g \in G(D)$ , we have that  $G(D) \cap D = \emptyset$ . It follows from (13) that the conditions of [Casau et al., 2017b, Theorem 1] are satisfied, thus  $\mathcal{A}$  is globally exponentially stable in the  $t$ -direction for the hybrid system (11).  $\square$

### 3.1 Construction of centrally synergistic potential function on $\mathbb{S}^n$

Given  $r \in \mathbb{S}^n$ , we construct a centrally synergistic potential function on  $\mathbb{S}^n$  using the height function:  $h_r(x) := 1 - r^\top x$  for all  $x \in \mathbb{S}^n$ . Let  $k > 0$ ,  $\mathcal{V} := \mathbb{S}^n \times \mathbb{S}^n \setminus \{(r, r)\}$ , and define  $V : \mathcal{V} \rightarrow \mathbb{R}$  for all  $(x, y) \in \mathcal{V}$  as

$$V(x, y) = \frac{h_r(x)}{h_r(x) + kh_y(x)} = \frac{1 - r^\top x}{1 - r^\top x + k(1 - y^\top x)}. \quad (14)$$

We now provide some differential properties of  $V$ , which follow from elementary calculation and some tedious manipulation, so they are presented without proof.

**Lemma 1** *The function  $V : \mathcal{V} \rightarrow [0, 1]$  defined in (14) satisfies*

$$\nabla V^y(x) = \frac{kV(x, y)y - (1 - V(x, y))r}{1 - r^\top x + k(1 - y^\top x)} \quad (15a)$$

$$|\Pi(x)\nabla V^y(x)|^2 = \frac{2kV(x, y)(1 - V(x, y))(1 - r^\top y)}{(1 - r^\top x + k(1 - y^\top x))^2} \quad (15b)$$

Given  $\gamma \in \mathbb{R}$  satisfying  $-1 < \gamma < 1$ , we define the set  $\mathcal{Q} \subset \mathbb{S}^n$  as

$$\mathcal{Q} = \{y \in \mathbb{S}^n : r^\top y \leq \gamma\}. \quad (16)$$

The boundary of  $\mathcal{Q}$  is  $\partial\mathcal{Q} = \{y \in \mathbb{S}^n : r^\top y = \gamma\}$ .

**Lemma 2** *Given  $r \in \mathbb{S}^n$  and  $\gamma \in [-1, 1]$ , let  $\mathcal{Q}$  be given by (16). The following holds for the function  $V$  given*

in (14)

$$\varrho(x) = \begin{cases} \mathcal{Q} & \text{if } x = r \\ -x & \text{if } r^\top x \geq -\gamma \\ \frac{\sigma(r^\top x)\Pi(x)r}{|\Pi(x)r|} + \alpha(r^\top x)x & \text{if } -1 < r^\top x < -\gamma \\ \partial\mathcal{Q} & \text{if } r^\top x = -1. \end{cases} \quad (17a)$$

$$\nu(x) = \begin{cases} \frac{1 - r^\top x}{1 - r^\top x + 2k} & \text{if } r^\top x \geq -\gamma \\ \frac{1 - r^\top x}{1 - r^\top x + k(1 - \alpha(r^\top x))} & \text{if } r^\top x < -\gamma \end{cases} \quad (17b)$$

for each  $x \in \mathbb{S}^n$ , where

$$\alpha(v) = \gamma v - \sqrt{(1 - v^2)(1 - \gamma^2)}, \quad (18a)$$

$$\sigma(v) = \gamma\sqrt{1 - v^2} + v\sqrt{1 - \gamma^2} \quad (18b)$$

for each  $v \in [-1, 1]$ , and  $\nu, \varrho$  are defined in (8) and (10), respectively.  $\square$

**PROOF.** Suppose that  $x = r$ . Then, according to (A.1) of Lemma A.1,  $V(x, y) = 0$  for all  $y \in \mathcal{Q}$ . Thus, any  $y \in \mathcal{Q}$  attains the minimum of  $y \mapsto V(r, y) = 0$ . If  $r^\top x \geq -\gamma$ , it follows that  $-x \in \mathcal{Q}$ . In this case, (A.3) of Lemma A.1 yields that the unconstrained minimizer of  $y \mapsto V(x, y)$ , which is  $-x$ , is also the minimizer of  $y \mapsto V(x, y)$  when  $y$  is constrained to  $y \in \mathcal{Q}$ . When  $x = -r$ , we have that  $V(-r, y) = 2/(2 + k(1 + r^\top y))$ . Clearly,  $y \mapsto V(-r, y)$  is minimized by maximizing  $r^\top y$ . When  $y$  is constrained to  $\mathcal{Q}$ , the maximum value of  $r^\top y$  is — by definition of  $\mathcal{Q}$  —  $\gamma$ , which is attained by any  $y$  satisfying  $r^\top y = \gamma$ , or equivalently,  $y \in \partial\mathcal{Q}$ . We now examine the case when  $-1 < r^\top x < -\gamma$ . Since  $-x \notin \mathcal{Q}$ , it suffices to study the solutions to the constrained minimization problem

$$\begin{aligned} & \underset{y}{\text{minimize}} && V(x, y) \\ & \text{subject to} && 1 - y^\top y = 0 \\ & && r^\top y - \gamma = 0. \end{aligned} \quad (19)$$

by means of Lagrange multipliers. Note that the constraints of the optimization problem are equivalent to  $y \in \partial\mathcal{Q}$ . Recalling (A.5), note that solutions to (19) are equivalently found as solutions to the constrained minimization problem

$$\begin{aligned} & \underset{y}{\text{minimize}} && y^\top x \\ & \text{subject to} && 1 - y^\top y = 0 \\ & && r^\top y - \gamma = 0. \end{aligned}$$

We proceed by constructing the Lagrangian,  $L : \mathbb{R}^{n+1} \times$

$\mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$  for each  $(y, \lambda, \beta) \in \mathbb{R}^{n+1} \times \mathbb{R} \times \mathbb{R}$  as

$$L(y, \lambda, \beta) = y^\top x + \lambda(y^\top r - \gamma) + \frac{1}{2}\beta(1 - y^\top y).$$

Setting  $\nabla L(y, \lambda, \beta) = 0$  yields

$$\beta y = x + \lambda r,$$

in addition to the constraint that  $y \in \mathcal{Q}$ . Seeing that  $\beta = 0$  implies that  $x \in \{r, -r\}$  (where we have already solved the problem at hand), we henceforth assume that  $\beta \neq 0$  and look for constants  $a, b \in \mathbb{R}$  such that

$$y = ax + br. \quad (20)$$

To ease notation, we let  $v = r^\top x$  for the remainder of the proof. Taking the norm of both sides of (20) and then noting that  $|y| = |x| = |r| = 1$ , yields

$$1 = a^2 + 2abv + b^2. \quad (21)$$

Then, by multiplying (20) on the left by  $r^\top$  and applying the constraint that  $r^\top y = \gamma$ , we arrive at

$$\gamma = av + b. \quad (22)$$

Solving (22) for  $b$  and substituting the result into (21) yields (with some rearrangement that

$$a^2 = \frac{1 - \gamma^2}{1 - v^2}.$$

By multiplying (20) on the left by  $x^\top$ , we see that

$$x^\top y = a(1 - v^2) + \gamma v.$$

Since  $V(x, y)$  is minimized when  $x^\top y$  is minimized and  $1 - v^2 = 1 - (r^\top x)^2 \geq 0$ , it follows that  $V$  is minimized when

$$a = -\sqrt{\frac{1 - \gamma^2}{1 - v^2}}.$$

By (22),

$$b = \gamma + v\sqrt{\frac{1 - \gamma^2}{1 - v^2}}.$$

Now, let

$$\begin{aligned} \sigma &= b\sqrt{1 - v^2} \\ &= \gamma\sqrt{1 - v^2} + v\sqrt{1 - \gamma^2} \\ \alpha &= a + vb \\ &= v\gamma - \sqrt{(1 - \gamma^2)(1 - v^2)} \end{aligned}$$

Note that  $|\Pi(x)r| = |(I - xx^\top)r| = \sqrt{1 - (r^\top x)^2}$ .

Then, by some rearrangement, it follows that

$$\begin{aligned} y &= (\alpha - vb)x + br \\ &= \alpha x + b(I - xx^\top)r \\ &= \alpha x + \sigma \frac{\Pi(x)r}{|\Pi(x)r|}. \end{aligned}$$

Recalling the definitions of  $\alpha$  and  $\sigma$  in (18a) and (18b), respectively, yields the desired result.  $\square$

Moreover, from Corollary A.1 and from definition (7) it follows

$$\mathcal{E}(V) = \{(x, x) \in \mathbb{S}^n \times \mathbb{S}^n : x \in \mathcal{Q}\},$$

for  $V$  given by (14). In the sequel, let

$$\mathcal{X} := \mathbb{S}^n \times \mathcal{Q},$$

denote the reduced state space and let the function  $\Delta : \mathbb{S}^n \rightarrow \mathbb{R}$  be given by

$$\Delta(x) := 1 - \nu(x).$$

for all  $x \in \mathbb{S}^n$ . We now note that both  $\mu$  and  $\Delta$  are continuous on  $\mathcal{X}$  and  $\mathbb{S}^n$ , respectively, and that  $\Delta$  agrees with  $\mu$  on the set  $\mathcal{E}$  (when the duplicated argument is ignored).

**Theorem 3** *The functions  $\mu : \mathcal{X} \rightarrow \mathbb{R}$  and  $\Delta : \mathbb{S}^n \rightarrow \mathbb{R}$  are continuous and for each  $x \in \mathcal{Q}$  (and therefore each  $(x, x) \in \mathcal{E}(V)$ ),*

$$\mu(x, x) = \Delta(x).$$

*Furthermore,  $\Delta(x) > 0$  for all  $x \in \mathbb{S}^n$ , and in particular,*

$$\min_{x \in \mathcal{Q}} \Delta(x) = \min_{x \in \mathbb{S}^n} \Delta(x) = \frac{1 + \gamma}{2/k + 1 + \gamma}$$

$\square$

**PROOF.** The continuity of  $\mu$  and  $\Delta$  follow from the continuity of  $\nu$ , which follows from the explicit calculation in (17b) of Lemma 2. Recalling that  $V(x, x) = 1$  for all  $(x, x) \in \mathcal{V}$  from (A.2) of Lemma A.1 and the calculation of  $\nu$  in (17b), we calculate  $\mu(x, x)$  for each  $x \in \mathcal{Q}$  as

$$\begin{aligned} \mu(x, x) &= V(x, x) - \nu(x) \\ &= 1 - \nu(x). \end{aligned}$$

We now show that  $\Delta(x) > 0$  for all  $x \in \mathbb{S}^n$  by finding an

explicit lower bound. Let

$$\begin{aligned}\eta_1(v) &= \frac{1-v}{1-v+2k} \\ \eta_2(v) &= \frac{1-v}{1-v+k(1-\alpha(v))} \\ \eta_3 &= \frac{2}{2+k(1+\gamma)}.\end{aligned}$$

It follows from (17a) and (17b) that

$$\min_{x \in \mathcal{Q}} \Delta(x) = 1 - \max \left\{ \max_{|v| \leq \gamma} \eta_1(v), \max_{-1 \leq v \leq -\gamma} \eta_2(v), \eta_3 \right\}$$

First, we calculate the derivative of  $\eta_1$  as

$$\eta_1'(v) = \frac{-2k}{(1-v+2k)^2}.$$

Since the denominator of  $\eta_1'(v)$  above is positive for all  $v \in [-1, 1]$  and because  $k > 0$ ,  $\eta_1'(v)$  is clearly negative for all  $v \in [-1, 1]$ . Hence,

$$\max_{|v| \leq \gamma} \eta_1(v) = \eta_1(-\gamma) = \frac{1+\gamma}{1+\gamma+2k}$$

After some effort, the derivative of  $\eta_2$  can be calculated as

$$\eta_2'(v) = k \frac{(1-v)\alpha'(v) - 1 + \alpha(v)}{(1-v+k(1-\alpha(v)))^2}.$$

Since the denominator of  $\eta_2'(v)$  above is always positive for  $v \in [-1, 1]$  and  $k > 0$ , the sign of  $\eta_2'(v)$  can be determined from the sign of the numerator. In this direction, we have

$$\begin{aligned}(1-v)\alpha'(v) - 1 + \alpha(v) \\ = \gamma - 1 - \sqrt{(1-v^2)(1-\gamma^2)} - v(1-v) \frac{\sqrt{1-\gamma^2}}{\sqrt{1-v^2}}.\end{aligned}$$

We note the equivalent series of inequalities (for all  $v \in (-1, 1)$ ),

$$\begin{aligned}\gamma - 1 - \sqrt{(1-v^2)(1-\gamma^2)} - v(1-v) \frac{\sqrt{1-\gamma^2}}{\sqrt{1-v^2}} &< 0 \\ (1-\gamma)\sqrt{1-v^2} + (1-v^2)\sqrt{1-\gamma^2} \\ + v(1-v)\sqrt{1-\gamma^2} &> 0 \\ (1-\gamma)\sqrt{1-v^2} + (1-v)\sqrt{1-\gamma^2} &> 0.\end{aligned}$$

Since  $|\gamma| < 1$  and  $|v| < 1$ , it follows that the inequality above is satisfied for all  $v \in (-1, 1)$ , so  $\eta_2'(v)$  is negative for all  $v \in (-1, -\gamma)$ . Noting that  $\alpha(-\gamma) = -1$  and  $\alpha(-1) = -\gamma$ , we have

$$\min_{-1 \leq v \leq -\gamma} \eta_2(v) = \eta_2(-\gamma) = \eta_1(-\gamma) = \frac{1+\gamma}{1+\gamma+2k}$$

and

$$\max_{-1 \leq v \leq -\gamma} \eta_2(v) = \eta_2(-1) = \eta_3 = \frac{2}{2+k(1+\gamma)}.$$

Thus,

$$\min_{x \in \mathcal{Q}} \Delta(x) = 1 - \frac{2}{2+k(1+\gamma)} = \frac{1+\gamma}{2/k+1+\gamma}.$$

□

Using (17b), we may compute (6), from which the next result follows.

**Corollary 1** *For any given  $r \in \mathbb{S}^n$  and  $\gamma \in (-1, 1)$ , the function (14) is a centrally synergistic potential function relative to  $r$  with synergy gap exceeding  $\delta$ , for any  $\delta \in (0, (1+\gamma)/(2/k+1+\gamma))$ .*

**PROOF.** This result follows from the fact that

$$\begin{aligned}\min\{\mu(x, y) : (x, y) \in \text{crit } V^y \setminus \{r\} \times \mathcal{Q}\} \\ = \min\{\mu(x, x) : x \in \mathcal{Q}\} \\ = \min\{1 - \nu(x) : x \in \mathcal{Q}\} \\ = \frac{1+\gamma}{2/k+1+\gamma}.\end{aligned}$$

In addition to global asymptotic stability of  $\mathcal{A}$  for (11), we also show below that the function (14) satisfies (13), thus it follows from Theorem 2 that global  $\mathcal{A}$  is also globally exponentially stable in the  $t$ -direction for (11). It follows from the fact that  $\mathcal{E}(V) = V^{-1}(1)$  (c.f. Lemma A.1) and from  $\mathcal{E}(V) \subset D$ , with  $D$  given in (9), that  $V^* \in \mathbb{R}$ , given by  $V^* = \max_{(x,y) \in C} V(x, y)$ , with  $C$  given in (9) satisfies  $0 < V^* < 1$ . We note that  $V^*$  exists since  $V$  is continuous on the compact set  $C$ . The next theorem follows naturally from these considerations

**Theorem 4** *The function  $V \in C^1(\mathbb{S}^n \times \mathcal{Q})$  given in (14) satisfies (13) with*

$$\underline{\alpha} = (2(1+k+\sqrt{1+2k\gamma+k^2}))^{-1}, \quad (23a)$$

$$\overline{\alpha} = (2(1+k-\sqrt{1+2k\gamma+k^2}))^{-1} \quad (23b)$$

$$\lambda = 2k(1-V^*)(1-\gamma) \left(1+k+\sqrt{1+2k\gamma+k^2}\right)^{-2}. \quad (23c)$$

*Therefore, the set (4) is globally exponentially stable in the  $t$ -direction for (11).*

**PROOF.** Since  $h_r(x) = 1 - r^\top x = |x - r|^2/2$  for every  $x \in \mathbb{S}^n$ , from (A.7) that (14) satisfies (13a) with  $\underline{\alpha}$  and



$\bar{\alpha}$  given by (23a) and (23b), respectively. Since  $r^\top y \leq \gamma$  for each  $(x, y) \in \mathbb{S}^n \times \mathcal{Q}$ , then  $1 - r^\top y \geq 1 - \gamma$ . It follows from the fact that  $V(x, y) \leq V^*$  for each  $(x, y) \in C$  and from Lemma A.2 that

$$|\Pi(x) \nabla V^y(x)|^2 \geq \frac{2k(1 - V^*)(1 - \gamma)}{(1 + k + \sqrt{1 + 2k\gamma + k^2})^2} V(x, y),$$

which proves that (13b) is satisfied. Global exponential stability of (4) for (11) follows from Theorem 2.  $\square$

### 3.2 Minimal Geodesics

We have shown that the proposed synergistic potential function (14) ensures not only global asymptotic stability of  $\mathcal{A}$  for (11), but also global exponential stability. Moreover, there is a third property of this function that is worth mentioning: for each  $x_0 \in D$ , the flows generated by the gradient vector field  $\Pi(x) \nabla V^{\varrho(x_0)}(x)$  converge to  $r$  through the minimal geodesic, i.e., the path of minimum distance between  $x_0$  and  $r$ , as proved next. For every pair of points  $p, q \in M$ , where  $M$  is a compact manifold, there is  $c : \mathbb{R} \rightarrow M$  with  $c(a) = p$  and  $c(b) = q$  for some  $a, b \in \mathbb{R}$  such that the length of  $c$ , given by  $L(c) = \int_a^b \left| \frac{dc}{dt}(\tau) \right| d\tau$ , is the lowest among all other paths with endpoints (c.f. Burns and Gidea [2005]). If a path  $p : [0, \infty) \rightarrow M$  is Lebesgue integrable, then its length is given by  $L^\infty(p) = \int_0^\infty \left| \frac{dp}{dt}(\tau) \right| d\tau$ . Considering the Riemannian manifold  $\mathbb{S}^n$  with the standard Euclidean metric, its (unit-velocity) geodesics are of the form  $x(t) = a \cos t + b \sin t$ , for each  $t \in \mathbb{R}$ , with  $a, b \in \mathbb{S}^n$  satisfying  $\langle a, b \rangle = 0$ , as shown in [Petersen, 2006, Example 30]. Given  $r \in \mathbb{S}^n$ , one verifies that the minimal geodesic between  $r$  and any given point  $x_0 \in \mathbb{S}^n \setminus \{r, -r\}$  is given by  $c(t) = x_0 \cos t + \Pi(x_0)r \sin t / |\Pi(x_0)r|$ , for each  $t \in [0, \arccos(x_0^\top r)]$ . If  $x_0 = -r$ , then each minimal geodesic from  $x_0$  to  $r$  is given by  $c(t) = x_0 \cos t + x^\perp \sin t$ , for each  $t \in [0, \pi]$ , where  $x^\perp \in \{y \in \mathbb{S}^n : \langle y, x_0 \rangle = 0\}$ . In particular, if  $x_0$  is antipodal to  $r$  and  $n > 1$  then there are uncountably many minimal geodesics from  $x_0$  to  $r$ .

It is possible to verify that, for every solution  $(x, y)$  of (11) and each  $(t, j), (t, j+1) \in \text{dom}(x, y)$ , the following holds:  $x(t, j) = x(t, j+1)$ . Therefore,  $x_{\downarrow t}(t) := x(t, J(t))$ , is defined for each  $t \in [0, \sup_t \text{dom}(x, y))$ , for each solution  $(x, y)$  to (11), where with  $J(t) := \max\{j : (t, j) \in \text{dom}(x, y)\}$ . Moreover, it is absolutely continuous, hence  $L^\infty(x_{\downarrow t})$  is well-defined. Choosing  $V$  in (14), we show next that the solutions to the hybrid system (11) that start in the jump set have minimal length.

**Lemma 3** *Consider the function  $V \in C^1(\mathbb{S}^n \times \mathcal{Q})$  given in (14). For each solution  $(x, y)$  to (11) with initial condition  $(x_0, y_0) \in D$ , we have that  $L^\infty(x_t) = L(c_{x_0, r})$ , where  $c_{x_0, r}$  denotes the minimal geodesic from  $x_0$  to  $r$ .*

**PROOF.** This result follows from the fact that the vec-

tor field  $\Pi(x) \nabla V^{\varrho(x_0)}(x)$  is tangent to the geodesic connecting  $r$  and  $x_0$  and from Theorem 4 which shows that solutions converge to  $r$ . Firstly, notice that, for each  $a, b \in \mathbb{S}^n$  satisfying  $\langle a, b \rangle = 0$ , we have that the charts  $(c_\pm^{-1}, O_\pm)$ , given by

$$\begin{aligned} c_+ &:= a \cos t + b \sin t & O_+ &:= c_+(\mathcal{I}), \\ c_- &:= -a \cos t + b \sin t & O_- &:= c_-(\mathcal{I}), \end{aligned}$$

where  $\mathcal{I} := (-\pi, \pi)$ , define a one-dimensional submanifold of  $\mathbb{S}^n$ , which we denote by  $\mathbb{S}^1(a, b)$ . Moreover, for each  $r \in \mathbb{S}^n$  and for each  $x_0 \in \mathbb{S}^n \setminus \{r, -r\}$ , the maximal geodesic through  $x_0$  and  $r$ , given by

$$c(t) = x_0 \cos t + \frac{\Pi(x_0)r}{|\Pi(x_0)r|} \sin t, \quad (24)$$

and defined for each  $t \in \mathbb{R}$ , belongs to  $\mathbb{S}^1(x_0, \frac{\Pi(x_0)r}{|\Pi(x_0)r|})$ , and the map (24) is a smooth immersion because its push-forward does not vanish for any  $t \in \mathbb{R}$ . Since the initial condition of the solution  $(x, y)$  belongs to the jump set, we have that  $(0, 1) \in \text{dom}(x, y)$ , and  $(x, y)(0, 1) = (x_0, \varrho(x_0))$ . The desired result follows from: the vector field  $\Pi(x) \nabla V^{\varrho(x_0)}(x)$  is tangent to  $\mathbb{S}^1(x_0, \frac{\Pi(x_0)r}{|\Pi(x_0)r|})$ ; from Theorem 4 we have that solutions converge to  $r$ . The former condition is verified if  $\nabla V^{\varrho(x_0)}(x)$  can be written as a linear combination of  $x_0$  and  $\frac{\Pi(x_0)r}{|\Pi(x_0)r|}$  for each  $x \in \mathbb{S}^1(x_0, \frac{\Pi(x_0)r}{|\Pi(x_0)r|})$ , because, in that case  $\Pi(x) \nabla V^{\varrho(x_0)}(x) \in \mathbb{T}_x \mathbb{S}^1(x_0, \frac{\Pi(x_0)r}{|\Pi(x_0)r|})$ . It is possible to rewrite (15a) as follows  $\nabla V^y(x) = k \frac{y(1-r^\top x) - r(1-x^\top y)}{(1-r^\top x + k(1-y^\top x))^2}$ , thus  $\nabla V^y(x)$  is a linear combination of  $y$  and  $r$ . Notice that  $r$  can be written as a linear combination of  $x_0$  and  $\frac{\Pi(x_0)r}{|\Pi(x_0)r|}$  because  $r = \Pi(x_0)r + x_0 x_0^\top r$ . On the other hand, from the definition of  $\varrho$  given in (17a), we have that

$$\varrho(x_0) = \begin{cases} \mathcal{Q} & \text{if } x_0 = r \\ -x_0 & \text{if } r^\top x_0 \geq -\gamma \\ \rho(x_0) & \text{if } -1 < r^\top x_0 < -\gamma \\ \partial \mathcal{Q} & \text{if } r^\top x_0 = -1. \end{cases}$$

with  $\varrho(x_0) := \sigma(r^\top x_0) \frac{\Pi(x_0)r}{|\Pi(x_0)r|} + \alpha(r^\top x_0) x_0$  for each  $x_0 \neq \pm r$ .  $\varrho(x_0)$  is also a linear combination of  $x_0$  and  $\frac{\Pi(x_0)r}{|\Pi(x_0)r|}$  for any  $x_0$  satisfying  $-1 < r^\top x_0$  because  $\alpha$  and  $\sigma$  are scalar maps. Also,  $r^\top x_0 \neq 1$  because in that case  $x_0 = r \notin D$ . If  $r^\top x_0 = -1$ , then any choice  $y \in \partial \mathcal{Q}$  satisfies the required condition for some geodesic.  $\square$

Note that function  $h_r$  induces a gradient-vector field that generates flows along geodesics for almost all initial conditions. However, unlike the aforementioned continuous feedback law, the controller presented in this section exponentially renders a reference point globally asymptotically stable, which is not possible with continuous feedback. Moreover, if we resolve the ambiguity at  $x = -r$

by means of some discontinuity, we still are left without any guarantees of robustness to small measurement noise. Since the hybrid controller presented in this section satisfies the hybrid basic conditions [Goebel et al., 2012, Assumption 6.5], the property of global asymptotic stability of  $\mathcal{A}$  for (11) is endowed with robustness to small measurement noise, as discussed in Goebel et al. [2012].

#### 4 Application to Trajectory Tracking for a Quadrotor

In this section, we apply the controller proposed in Section 3 to the problem of trajectory tracking for a quadrotor vehicle, i.e., an aerial vehicle with four counter rotating rotors that are aligned with a direction which is fixed relative to the body of the vehicle, as described in Hamel et al. [2002]. The dynamics of a thrust vectored vehicle such as a quadrotor can be described by

$$\dot{p} = v \quad (25a)$$

$$\dot{v} = Rru + g \quad (25b)$$

$$\dot{R} = RS(\omega) \quad (25c)$$

where  $p \in \mathbb{R}^3$  and  $v \in \mathbb{R}^3$  denote the position and the velocity of the vehicle with respect to the inertial reference frame (in inertial coordinates),  $R \in \text{SO}(3) := \{R \in \mathbb{R}^{3 \times 3} : R^\top R = I_3, \det(R) = 1\}$  is the rotation matrix that maps vectors in body-fixed coordinates to inertial coordinates,  $g \in \mathbb{R}^3$  represents the gravity vector and  $r \in \mathbb{S}^2 := \{x \in \mathbb{R}^3 : |x| = 1\}$  is the thrust vector in body-fixed coordinates. Furthermore, the inputs to (25) are  $\omega \in \mathbb{R}^3$  and  $u \in \mathbb{R}$  which represent the angular velocity in body-fixed coordinates and the magnitude of the thrust, respectively. The dynamical model (25) is a simplification of the one provided in Hamel et al. [2002] that better suits our experimental setup, since there the *Blade 200 QX* quadrotor that is used in the experiments has an embedded controller that tracks angular velocity and thrust commands. Furthermore, we assume that the reference trajectory satisfies the following assumption.

**Assumption 1** *The reference trajectory  $t \mapsto p_d(t)$  is defined for each  $t \geq 0$  and there exist  $M_2 \in (0, |g|)$  and  $M_3 > 0$  such that  $|\ddot{p}_d(t)| \leq M_2$  and  $p_d^{(3)}(t) \leq M_3$  for all  $t \geq 0$ .*

Given a path that satisfies Assumption 1, we define the tracking errors as  $\tilde{p} := p - p_d$  and  $\tilde{v} := v - \dot{p}_d$ , whose dynamics can be derived from (25) and are given by:

$$\dot{\tilde{p}} = \tilde{v}, \quad \dot{\tilde{v}} = Rru + g - \ddot{p}_d. \quad (26)$$

Given  $w : \mathbb{R}^6 \rightarrow \mathbb{R}^3$ , we define

$$\rho(z) := \frac{w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d}{|w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d|} \quad (27a)$$

$$\kappa_u(z, R) := r^\top R^\top (w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d) \quad (27b)$$

for each  $z := (\ddot{p}_d, \tilde{p}, \tilde{v}) \in M_2 \mathbb{B} \times \mathbb{R}^6$  and  $R \in \text{SO}(3)$ . Note that, if  $Rr = \rho(z)$  and  $u = \kappa_u(z, R)$ , we obtain

$$\dot{\tilde{p}} = \tilde{v}, \quad \dot{\tilde{v}} = w(\tilde{p}, \tilde{v}) \quad (28)$$

from (26), provided that

$$w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d \neq 0. \quad (29)$$

On the other hand, if  $Rr \neq \rho(z)$ , then  $\kappa_u(z, R)$  is the solution to the least-squares problem:

$$\min\{|Rru + g - \ddot{p}_d - w(\tilde{p}, \tilde{v})|^2 : u \in \mathbb{R}\}.$$

The mismatch between  $Rr$  and  $\rho(z)$  may lead to an increase in the the position and velocity tracking errors until the thrust vector  $Rr$  is aligned with  $\rho(z)$ . The controller design is a two-step process, where we start by designing a feedback law  $(\tilde{p}, \tilde{v}) \mapsto w(\tilde{p}, \tilde{v})$  that exponentially stabilizes the origin of (28) and then we design a partial attitude tracking controller that exponentially stabilizes  $\rho(z)$  for the dynamics of the thrust vector  $Rr$ .

##### 4.1 Controller for the Position Subsystem

We show next that it is possible to satisfy (29) and exponentially stabilize the origin of (28) from an arbitrary compact set  $U$  using a saturated linear feedback law

$$w(\tilde{p}, \tilde{v}) := \text{Sat}_b \left( K \begin{bmatrix} \tilde{p}^\top & \tilde{v}^\top \end{bmatrix}^\top \right) \quad (30)$$

where  $b \in (0, |g| - M_2)$  and

$$v \mapsto \text{Sat}_b(v) := \begin{bmatrix} \text{sat}_b(v_1) & \dots & \text{sat}_b(v_n) \end{bmatrix}^\top$$

for each  $v \in \mathbb{R}^n$  with  $\text{sat}_b : \mathbb{R} \rightarrow \mathbb{R}$  continuously differentiable and nondecreasing, satisfying  $\text{sat}_b(x) = x$  for all  $x \in [-b, b]$  and  $\text{sat}_b(x) \in (M_2 - |g|, |g| - M_2)$  for each  $x \in \mathbb{R}$ . Defining  $\Omega_P(\ell) := \{x \in \mathbb{R}^n : x^\top P x \leq \ell\}$  for  $P \in \mathbb{R}^{n \times n}$  and  $\ell > 0$ , if we select  $(H, \ell_H) \in \mathbb{S}_{>0}^6$  such that  $\Omega_H(\ell_H)$  is a bounding ellipsoid for  $U$ , then we guarantee that the bounds  $\pm b$  of the saturation function (30) are not reached if there exists  $(P, \ell_P) \in \mathbb{S}_{>0}^6 \times \mathbb{R}_{>0}$  and  $K \in \mathbb{R}^{3 \times 6}$ , such that  $\Omega_P(\ell_P)$  is forward invariant for every solution to (28) from  $U$  and

$$U \subset \Omega_H(\ell_H) \subset \Omega_P(\ell_P) \subset \Omega_{K^\top K}(b^2). \quad (31)$$

For each compact set  $U \subset \mathbb{R}^6$ , it is possible to select controller parameters that satisfy (31) as well as

$$(A + BK)^\top P + P(A + BK) \preceq -\hat{Q} - K^\top \hat{R} K \quad (32)$$

where

$$A := \begin{bmatrix} 0 & I_3 \\ 0 & 0 \end{bmatrix}, \quad B := \begin{bmatrix} 0 \\ I_3 \end{bmatrix}.$$

thus guaranteeing also the exponential stability of the origin of (28) from  $U$ , as stated in the proposition below.

**Proposition 1** *For each compact set  $U \subset \mathbb{R}^6$ ,  $b > 0$ ,  $\ell_P > 0$ ,  $\hat{R} \in \mathbb{S}_{>0}^3$  and  $H \in \mathbb{S}_{>0}^6$ , there exists  $K \in \mathbb{R}^{3 \times 6}$  and  $\hat{Q}, P \in \mathbb{S}_{>0}^6$  such that the conditions (31) and (32) are satisfied.*

**PROOF.** From [Saber et al., 2012, Lemma 4.20], it follows that there is a function  $\epsilon \mapsto \hat{Q}(\epsilon)$ , with the properties

$$\hat{Q}(\epsilon) \succ 0 \quad \text{and} \quad \frac{d\hat{Q}(\epsilon)}{d\epsilon} \succ 0$$

for each  $\epsilon \in (0, 1]$ , that generates a unique solution  $P(\epsilon) \succ 0$  to

$$A^\top P(\epsilon) + P(\epsilon)A - P(\epsilon)B\hat{R}^{-1}B^\top P(\epsilon) + \hat{Q}(\epsilon) = 0 \quad (33)$$

satisfying  $P(\epsilon) \rightarrow 0$  as  $\epsilon \rightarrow 0$ .

Choosing  $K = -\hat{R}^{-1}B^\top P(\epsilon)$ , it follows that

$$(A + BK)^\top P(\epsilon) + P(\epsilon)(A + BK) = A^\top P(\epsilon) + P(\epsilon)A - 2P(\epsilon)B\hat{R}^{-1}B^\top P(\epsilon). \quad (34)$$

Adding and subtracting  $\hat{Q}(\epsilon)$  to the right hand side of (34) and using (33) yields

$$A^\top P(\epsilon) + P(\epsilon)A + K^\top B^\top P(\epsilon) + P(\epsilon)BK = -P(\epsilon)B\hat{R}^{-1}B^\top P(\epsilon) - \hat{Q}(\epsilon).$$

Therefore, the condition (32) is satisfied with  $P = P(\epsilon)$  and  $\hat{Q} = \hat{Q}(\epsilon)$ . The condition  $\Omega_H(\ell_H) \subset \Omega_P(\ell_P)$  is satisfied if and only if

$$\frac{P(\epsilon)}{\ell_P} \preceq \frac{H}{\ell_H}, \quad (35)$$

and the condition  $\Omega_P(\ell_P) \subset \Omega_{K^\top K}(b^2)$  is satisfied if and only if  $\frac{K^\top K}{b^2} \preceq \frac{P(\epsilon)}{\ell_P} \iff P(\epsilon)^{\frac{1}{2}} B \hat{R}^{-2} B^\top P(\epsilon)^{\frac{1}{2}} \preceq \frac{b^2}{\ell_P} I_6$ , which is equivalent to

$$\left| \hat{R}^{-1} B^\top P(\epsilon)^{\frac{1}{2}} v \right|^2 \leq \frac{b^2}{\ell_P} |v|^2 \quad \forall v \in \mathbb{R}^6. \quad (36)$$

From [Bernstein, 2009, Proposition 9.4.9], it follows that (36) is satisfied if

$$\sigma_{\max}(\hat{R}^{-1} B^\top P(\epsilon)^{\frac{1}{2}})^2 \leq \frac{b^2}{\ell_P} \quad (37)$$

holds. Since  $P(\epsilon)$  can be made arbitrarily close to zero by choosing a small enough  $\epsilon$ , it follows that, for each  $\ell_P > 0$ ,  $\ell_H > 0$ ,  $\hat{R} \succ 0$  and  $b > 0$ , it is possible to find  $P(\epsilon)$  satisfying both (37) and (35).  $\square$

The previous proposition is very important to the following theorem, which constitutes the main result of this section.

**Theorem 5** *For each  $b \in (0, |g| - M_2)$ , and each compact set  $U \subset \mathbb{R}^6$ , there exists  $K \in \mathbb{R}^{3 \times 6}$  such that the origin of the closed-loop system resulting from the interconnection between (28) and (30) is exponentially stable from  $U$ . Moreover, each solution to (28) from  $U$ , denoted by  $t \mapsto (\tilde{p}, \tilde{v})(t)$ , satisfies  $\left| K \begin{bmatrix} \tilde{p}(t)^\top & \tilde{v}(t)^\top \end{bmatrix}^\top \right| \leq b$  for each  $t \geq 0$ .*

**PROOF.** Choosing a positive definite matrix  $H \in \mathbb{S}_{>0}^6$  such that  $\Omega_H(1)$  is a bounding ellipsoid for  $U$ , it follows from Proposition 1 that there exists  $K \in \mathbb{R}^{3 \times 6}$  and  $\hat{Q}, P \in \mathbb{S}_{>0}^6$  such that the conditions (31) and (32) hold for any positive definite matrix  $\hat{R} \in \mathbb{R}^{3 \times 3}$  and any  $\ell_P > 0$ . Let  $V_p(\tilde{p}, \tilde{v}) := \begin{bmatrix} \tilde{p}^\top & \tilde{v}^\top \end{bmatrix} P \begin{bmatrix} \tilde{p}^\top & \tilde{v}^\top \end{bmatrix}^\top$ , for each  $(\tilde{p}, \tilde{v}) \in \mathbb{R}^6$ , which is a positive definite function relative to  $\{(\tilde{p}, \tilde{v}) \in \mathbb{R}^6 : \tilde{p} = \tilde{v} = 0\}$  and satisfies: for each  $(\tilde{p}, \tilde{v}) \in \Omega_P(\ell_P)$

$$\begin{aligned} & \langle \nabla V_p(\tilde{p}, \tilde{v}), f_p(\tilde{p}, \tilde{v}) \rangle \\ & \leq - \begin{bmatrix} \tilde{p}^\top & \tilde{v}^\top \end{bmatrix} (\hat{Q} + K^\top \hat{R} K) \begin{bmatrix} \tilde{p}^\top & \tilde{v}^\top \end{bmatrix}^\top \end{aligned} \quad (38)$$

where  $f_p(\tilde{p}, \tilde{v}) := \begin{bmatrix} \tilde{v}^\top & \begin{bmatrix} \tilde{p}^\top & \tilde{v}^\top \end{bmatrix} K^\top \end{bmatrix}^\top$ . It follows from the conditions (31), that every solution  $t \mapsto (\tilde{p}, \tilde{v})(t)$  to (28) from  $U$  satisfies  $\left| K \begin{bmatrix} \tilde{p}(t)^\top \\ \tilde{v}(t)^\top \end{bmatrix} \right| \leq b < |g| - M_2$  for all  $t \geq 0$ .  $\square$

In the next lemma, we show that the controller gain  $K$  in (30) can be computed from an optimization problem that provides a sub-optimal solution to the  $H_2$ -minimization problem (c.f. [Scherer and Weiland, 2000, Proposition 3.11]).

**Lemma 4** *Given  $\ell_H > 0$ ,  $\ell_P > 0$ ,  $\hat{Q} \in \mathbb{S}_{>0}^6$  and  $\hat{R} \in \mathbb{S}_{>0}^3$ , there exists a solution  $P \in \mathbb{S}_{>0}^6$ ,  $K \in \mathbb{R}^{3 \times 6}$  to (31)*

and (32) if and only if there exists a solution  $Y \in \mathbb{S}_{>0}^6$  and  $L \in \mathbb{R}^{3 \times 6}$  to

$$\begin{aligned} & \text{minimize} \quad \text{trace}(Y^{-1}) \\ & \text{subject to} \quad (Y, L) \in \chi_{LMI} \end{aligned}$$

where  $\chi_{LMI}$  is the set of linear matrix inequalities:

$$\begin{bmatrix} -(AY + BL)^\top - (AY + BL) & Y & L^\top \\ Y & \hat{Q}^{-1} & 0 \\ L & 0 & \hat{R}^{-1} \end{bmatrix} \succeq 0 \quad (39a)$$

$$Y \succeq \frac{\ell_H}{\ell_P} H^{-1} \quad (39b)$$

$$\begin{bmatrix} Y & L \\ L^\top & \frac{b^2}{\ell_P} I_6 \end{bmatrix} \succeq 0 \quad (39c)$$

in which case  $P = Y^{-1}$  and  $K = LY^{-1}$ .

**PROOF.** Since  $\hat{Q} \succ 0$ , we can rewrite (32) as follows:

$$\begin{bmatrix} X - Y\hat{Q}Y - L^\top \hat{R}L & 0 \\ 0 & \hat{Q}^{-1} \end{bmatrix} \succeq 0 \quad (40)$$

where  $X := -(AY + BL)^\top - (AY + BL)$ . Equivalently, we obtain

$$\begin{bmatrix} X & Y \\ Y & \hat{Q}^{-1} \end{bmatrix} - \begin{bmatrix} L^\top \\ 0 \end{bmatrix} \hat{R} \begin{bmatrix} L & 0 \end{bmatrix} \succeq 0 \quad (41)$$

from (40) by means of a congruence transformation. The application of [Bernstein, 2009, Proposition 8.2.4.iii)] to (41) yields (39a). The condition (39b) follows from the fact that  $\Omega_H(\ell_H) \subset \Omega_P(\ell_P)$  if and only if  $P \preceq \frac{\ell_P}{\ell_H} H$ , using  $Y = P^{-1}$ . The condition (39c) follows from the fact that  $\Omega_P(\ell_P) \subset \Omega_{K^\top K}(b^2)$  if and only if  $\frac{K^\top K}{b^2} \preceq \frac{P}{\ell_P}$  and [Bernstein, 2009, Proposition 8.2.4] as follows:

$$\begin{aligned} \frac{K^\top K}{b^2} \preceq \frac{P}{\ell_P} & \iff \frac{YK^\top KY}{b^2} \preceq \frac{Y}{\ell_P} \\ & \iff \frac{Y}{\ell_P} - \frac{L^\top L}{b^2} \succeq 0 \end{aligned}$$

where we have, once again, used  $Y = P^{-1}$  and  $L = KY$ .

## 4.2 Partial attitude tracking

In this section, we develop a controller for (25) that tracks a reference trajectory satisfying Assumption 1. The desired acceleration, imposed by the reference trajectory upon the vehicle, is achieved by aligning the

thrust vector  $Rr$  with the direction of the desired acceleration. We refer to this as partial attitude tracking because we do not control rotations around the thrust vector. For controller design purposes, let us assume the following.

**Assumption 2** Given  $\delta > 0$ , there exists a function  $V \in \mathcal{Q}_r$  with synergy gap exceeding  $\delta$  that satisfies

$$\begin{aligned} \underline{\alpha} |x - r|^2 & \leq V(x, y) \leq \bar{\alpha} |x - r|^2 \quad \forall (x, y) \in C \cup D \\ |\Pi(x) \nabla V^y(x)|^2 & \geq \lambda_1 V(x, y) \quad \forall (x, y) \in C. \end{aligned}$$

for some  $\underline{\alpha}, \bar{\alpha}, \lambda_1 > 0$ , where

$$C := \{(x, y) \in \mathbb{S}^2 \times \mathcal{Q} : \mu(x, y) \leq \delta\}, \quad (42a)$$

$$D := \{(x, y) \in \mathbb{S}^2 \times \mathcal{Q} : \mu(x, y) \geq \delta\}. \quad (42b)$$

Under Assumption 2, the application of the controller developed in Section 3 yields the closed-loop system  $\mathcal{H}_1 := (C_1, F_1, D_1, G_1)$  with state  $\zeta := (z, R, y) \in \mathcal{Z} := M_2\mathbb{B} \times \mathbb{R}^6 \times \text{SO}(3) \times \mathbb{S}^2$ , given by

$$\begin{aligned} F_1(\zeta) &:= \{(F_p(p_d^{(3)}, \zeta), RS(\kappa_1(p_d^{(3)}, \zeta)), 0) : p_d^{(3)} \in M_3\mathbb{B} \\ & \quad \zeta \in C_1 := \{\zeta \in \mathcal{Z} : (R^\top \rho(z), y) \in C\} \\ G_1(\zeta) &:= (z, R, \varrho(R^\top \rho(z))) \\ & \quad \zeta \in D_1 := \{\zeta \in \mathcal{Z} : (R^\top \rho(z), y) \in D\} \end{aligned} \quad (43)$$

where  $F_p(p_d^{(3)}, \zeta) := (p_d^{(3)}, \tilde{v}, Rr\kappa_u(z, R) + g - \ddot{p}_d)$  for each  $(p_d^{(3)}, \zeta) \in M_3\mathbb{B} \times \mathcal{Z}$ , the inputs  $u$  and  $\omega$  were assigned to  $\kappa_u(z, \mathcal{R}(\zeta))$  and

$$\begin{aligned} \kappa_1(p_d^{(3)}, \zeta) &:= S(R^\top \rho(z))(R^\top \mathcal{D}_z(\rho(z)) F_p(p_d^{(3)}, \zeta) \\ & \quad + (k_1 + k_p \nu^*(z)) \nabla V^y(R^\top \rho(z))), \end{aligned} \quad (44)$$

for each  $(p_d^{(3)}, \zeta) \in M_3\mathbb{B} \times \mathcal{Z}$ , respectively, with  $k_1 > 0$ ,  $k_p > 0$  and

$$\nu^*(z) := \frac{2}{\sqrt{\underline{\alpha}}} \sigma_{\max} \left( \begin{bmatrix} 0 & I_3 \end{bmatrix} P^{\frac{1}{2}} \right) |w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d|, \quad (45)$$

for each  $z \in M_2\mathbb{B} \times \mathbb{R}^6$ . The closed-loop system represented by  $\mathcal{H}_1$  in (43) inherits the switching logic from (9) but not the same feedback law. The feedback law (44) is comprised of negative feedback of the attitude tracking error as in (9), but also a feedforward term that takes into account the fact that the reference we wish to track is not a constant. Moreover, the function  $\nu^*$  is used to increase the gain of the attitude controller as the position error increases. Given a reference trajectory satisfying Assumption 1, compact set of initial conditions  $U \subset \mathbb{R}^6$  for the position and velocity errors and a synergistic potential function satisfying Assumption 2, the controller design is as follows:

- (C1) Select  $H \in \mathbb{S}_{>0}^6$  such that  $\Omega_H(1)$  is a bounding ellipsoid for  $U$ ;

(C2) Select  $k_p, \bar{k}_1 > 0$  so that

$$k_p \bar{k}_1 \lambda_1 > 1 \quad (46)$$

where  $\lambda_1$  is given in Assumption 2;

(C3) Given  $b \in (0, |g| - M_2)$  and  $\hat{R} \in \mathbb{S}_{>0}^6$ , select  $\ell_H = \ell_P = (1 + \bar{\nu})^2$ , where

$$\bar{\nu} := \bar{k}_1 \sqrt{\max_{(x,y) \in \mathbb{S}^2 \times \mathcal{Q}} V(x,y)}$$

and  $\hat{Q} \in \mathbb{S}_{>0}^6$  small enough and so that the constraints (31) and (32) are feasible;

(C4) Compute the controller gain  $K$  by means of the optimization problem  $\min\{\text{trace}(P) : (P, K) \in \chi\}$  where  $\chi$  represents the constraints (31) and (32).

This controller design enables exponential stability, as proved next.

**Theorem 6** *Let Assumptions 1 and 2 hold. For each compact set  $U \subset \mathbb{R}^6$ , if (C1)–(C4) are satisfied, then  $\mathcal{A}_1 := \{\zeta \in \mathcal{Z} : \tilde{p} = \tilde{v} = 0, \rho(z) = Rr\}$  is exponentially stable in the  $t$ -direction from  $M_2\mathbb{B} \times U \times \text{SO}(3) \times \mathcal{Q}$  for the hybrid system (43).*

**PROOF.** Firstly, we prove that the hybrid system (43) satisfies the hybrid basic conditions. It follows from Proposition A.1 that  $\mu$  is continuous, thus both  $C_1$  and  $D_1$  are closed because they correspond to the inverse image of a closed set through a continuous map. It follows from  $b \in (0, |g| - M_2)$  and the properties of the saturation function that  $\mathcal{D}_z(\rho(z))$  is continuously differentiable for each  $z \in M_2\mathbb{B} \times \mathbb{R}^6$ , hence  $F_1$  is outer semicontinuous and locally bounded relative to  $C_1$ . It follows from the outer semicontinuity of  $\varrho$  that  $G_1$  is outer semicontinuous relative to  $D_1$  and, since  $\varrho$  takes values over a compact set,  $G_1$  is locally bounded relative to  $D_1$ . Let

$$W_1(\zeta) := \sqrt{V_p(\tilde{p}, \tilde{v})} + \bar{k}_1 \sqrt{V(x, y)} \quad \forall \zeta \in \mathcal{Z} \quad (47)$$

with  $P := Y^{-1}$  and  $x := R^\top \rho(z)$ . It follows from Assumption 2 that

$$\begin{aligned} \min\{\sqrt{\lambda_{\min}(P)}, \bar{k}_1 \sqrt{\underline{\alpha}}\} |(\tilde{p}, \tilde{v}, x - r)| &\leq W_1(\zeta) \\ &\leq \sqrt{\lambda_{\max}(P) + \bar{k}_1^2 \bar{\alpha}} |(\tilde{p}, \tilde{v}, x - r)|. \end{aligned}$$

for each  $\zeta \in C_1 \cup D_1$ . It follows from the assumptions

that the time derivative of (47) is given by

$$\begin{aligned} \langle W_1(\zeta), f_1 \rangle &= \\ &+ \frac{1}{2\sqrt{V_p(\tilde{p}, \tilde{v})}} \nabla V_p(\tilde{p}, \tilde{v})^\top \begin{bmatrix} \tilde{v} \\ Rr\kappa_u(z, R) + g - \ddot{p}_d \end{bmatrix} \\ &+ \frac{\bar{k}_1}{2\sqrt{V(x, y)}} \nabla V^y(x)^\top (S(x)\kappa_1(\zeta, y) \\ &- R^\top \mathcal{D}_z(\rho(z)) F_p(p_d^{(3)}, \zeta)) \end{aligned} \quad (48)$$

for each  $f_1 \in F_1(\zeta)$ ,  $\zeta \in \hat{\Omega} := \{\zeta \in \mathcal{Z} : W_1(\zeta) \leq 1 + \bar{\nu}\}$ , because  $W_1(\zeta) \leq 1 + \bar{\nu}$  implies that  $V_p(\tilde{p}, \tilde{v}) \leq (1 + \bar{\nu})^2$  and, by the construction (C1)–(C3), the saturation bound  $b$  is not exceeded in this set since it satisfies (31). Using  $xx^\top = S(x)^2 + I_3$  for each  $x \in \mathbb{S}^2$ , and replacing (27b) and (44) into (48) yields

$$\begin{aligned} W_1^o(\zeta; f_1) &= \frac{1}{2\sqrt{V_p(\tilde{p}, \tilde{v})}} \nabla V_p(\tilde{p}, \tilde{v})^\top \begin{bmatrix} \tilde{v} \\ w(\tilde{p}, \tilde{v}) \end{bmatrix} \\ &+ \frac{1}{2\sqrt{V_p(\tilde{p}, \tilde{v})}} \nabla V_p(\tilde{p}, \tilde{v})^\top \begin{bmatrix} 0 \\ I_3 \end{bmatrix} S(Rr)^2 (w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d) \\ &- \frac{\bar{k}_1(k_1 + k_p \nu^*(z))}{2\sqrt{V(x, y)}} |S(x)^2 \nabla V^y(x)|^2 \end{aligned} \quad (49)$$

for each  $f_1 \in F_1(\zeta)$ ,  $\zeta \in \hat{\Omega}$ . Using  $\nabla V_p(\tilde{p}, \tilde{v}) = 2P \begin{bmatrix} \tilde{p}^\top & \tilde{v}^\top \end{bmatrix}^\top$  and (38) it follows from (49) that

$$\begin{aligned} W_1^o(\zeta; f_1) &\leq -\frac{1}{2\sqrt{V_p(\tilde{p}, \tilde{v})}} \begin{bmatrix} \tilde{p} \\ \tilde{v} \end{bmatrix}^\top (\hat{Q} + K^\top \hat{R} K) \begin{bmatrix} \tilde{p} \\ \tilde{v} \end{bmatrix} \\ &+ \frac{1}{\sqrt{V_p(\tilde{p}, \tilde{v})}} \left| \begin{bmatrix} 0 & I_3 \end{bmatrix} P \begin{bmatrix} \tilde{p} \\ \tilde{v} \end{bmatrix} \right| |S(Rr)^2 (w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d)| \\ &- \frac{\bar{k}_1(k_1 + k_p \nu^*(z))}{2\sqrt{V(x, y)}} |S(x)^2 \nabla V^y(x)|^2 \end{aligned} \quad (50)$$

for each  $f_1 \in F_1(\zeta)$   $\zeta \in \hat{\Omega}$ . Since

$$|S(Rr)^2 (w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d)| \leq |x - r| |w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d|$$

for each  $\zeta \in C_1 \cup D_1$  and

$$\frac{\left| \begin{bmatrix} 0 & I_3 \end{bmatrix} P z \right|}{\sqrt{z^\top P z}} = \frac{\left| \begin{bmatrix} 0 & I_3 \end{bmatrix} P^{1/2} \tilde{z} \right|}{|\tilde{z}|} \leq \sigma_{\max} \left( \begin{bmatrix} 0 & I_3 \end{bmatrix} P^{1/2} \right)$$

for each  $z \in \mathbb{R}^6 \setminus \{0\}$  where  $\tilde{z} := P^{1/2} z$ , it follows (50)



that

$$\begin{aligned} W_1^o(\zeta; f_1) &\leq -\frac{\lambda_{\min}(\hat{Q} + K^\top \hat{R}K)}{2\sqrt{\lambda_{\max}(P)}} \sqrt{V_p(\tilde{p}, \tilde{v})} \\ &+ \sigma_{\max} \left( \begin{bmatrix} 0 & I_3 \end{bmatrix} P^{1/2} \right) |x - r| |w(\tilde{p}, \tilde{v}) - g + \ddot{p}_d| \\ &- \frac{\bar{k}_1(k_1 + k_p \nu^*(z))}{2\sqrt{V(x, y)}} |S(x)^2 \nabla V^y(x)|^2 \end{aligned} \quad (51)$$

for each  $f_1 \in F_1(\zeta)$ ,  $\zeta \in \hat{\Omega}$ . Replacing (45) into (51), we have that

$$\begin{aligned} W_1^o(\zeta; f_1) &\leq -\frac{\lambda_{\min}(\hat{Q} + K^\top \hat{R}K)}{2\sqrt{\lambda_{\max}(P)}} \sqrt{V_p(\tilde{p}, \tilde{v})} \\ &- \frac{\bar{k}_1(k_1 + k_p \nu^*(z))\lambda_1}{2} \sqrt{V(x, y)} \\ &+ \frac{\nu^*(z)}{2} \sqrt{V(x, y)} \end{aligned}$$

for each  $f_1 \in F_1(\zeta)$ ,  $\zeta \in \hat{\Omega}$ . It follows from (C2) that

$$\begin{aligned} W_1^o(\zeta; f_1) &\leq -\frac{\lambda_{\min}(\hat{Q} + K^\top \hat{R}K)}{2\sqrt{\lambda_{\max}(P)}} \sqrt{V_p(\tilde{p}, \tilde{v})} \\ &- \frac{\bar{k}_1 k_1 \lambda_1 \sqrt{V(x, y)}}{2} \end{aligned} \quad (52)$$

for each  $f_1 \in F_1(\zeta)$ ,  $\zeta \in \hat{\Omega} := \{\zeta \in \mathcal{Z} : W_1(\zeta) \leq 1 + \bar{\nu}\}$ . It follows from (31) that  $(\tilde{p}, \tilde{v}) \in \Omega_H(1)$  implies  $(\tilde{p}, \tilde{v}) \in \Omega_P(1)$ , hence, for each solution from  $M_2\mathbb{B} \times U \times \text{SO}(3) \times \mathcal{Q}$ , we have  $V_p(\tilde{p}(0, 0), \tilde{v}(0, 0)) \leq 1$  and, consequently,  $W_1(\zeta(0, 0)) \leq 1 + \bar{\nu}$ . From (46) and (52), we have  $W_1^o(\zeta; f_1) \leq -\lambda W_1(\zeta)$  for each  $f_1 \in F_1(\zeta)$ ,

$\zeta \in \hat{\Omega}$  with  $\lambda := \min \left\{ \frac{\lambda_{\min}(\hat{Q} + K^\top \hat{R}K)}{2\sqrt{\lambda_{\max}(P)}}, \frac{\bar{k}_1 k_1 \lambda_1}{2} \right\}$ . Note

that  $V$  is a synergistic potential function relative to  $r$  by assumption, thus it satisfies  $V(x, g_y) \leq V(x, y) - \delta$  for each  $g_y \in \rho(x)$  and  $\zeta \in D_1$  by construction, and  $G_1(D_1) \cap D_1 = \emptyset$  because  $\mu(x, g_y) = 0 < \delta$  for each  $g_y \in \rho(x)$  for each  $\zeta \in D_1$ . Each solution  $(t, j) \mapsto \zeta(t, j)$  to (43) from  $M_2\mathbb{B} \times U \times \text{SO}(3) \times \mathcal{Q}$  is such that the initial condition  $\zeta(0, 0)$  belongs to the compact set  $\hat{\Omega}$ . Since  $W_1$  is strictly decreasing during flows and jumps of (43), it follows that  $\hat{\Omega}$  is forward invariant for each solution from  $M_2\mathbb{B} \times U \times \text{SO}(3) \times \mathcal{Q}$ . Since  $G_1(D_1) \subset C_1 \cup D_1$  and  $F_1(\zeta)$  belongs to the tangent cone to  $C_1$  at  $\zeta$  for each  $\zeta \in C_1 \setminus D_1$ , it follows from [Goebel et al., 2012, Proposition 6.10] that each maximal solution to  $\mathcal{H}_1$  from  $M_2\mathbb{B} \times U \times \text{SO}(3) \times \mathcal{Q}$  is complete, hence  $\mathcal{A}_1$  is exponentially stable in the  $t$ -direction from  $M_2\mathbb{B} \times U \times \text{SO}(3) \times \mathcal{Q}$  for the hybrid system (43).  $\square$

It should be pointed out that, underlying the controller design described in (C1) through (C4), there is a trade-off to be resolved: if  $k_p \ll 1$ , then the position controller is going to have a low gain  $K$  which results in

large deviations from the reference; on the other hand, if controller gain  $K$  is large, the function  $\nu^*$  might increase the gain on the attitude controller beyond what is acceptable in practical terms.

#### 4.3 Simulation Results

$$\begin{aligned} p(0, 0) &:= p_d(0, 0), \quad v(0, 0) := v_d(0, 0), \\ y_1(0, 0) &= \begin{bmatrix} 0 & 0 & 1 \end{bmatrix}^\top, \quad R(0, 0) := \text{diag} \left( \begin{bmatrix} 1 & -1 & -1 \end{bmatrix} \right). \end{aligned}$$

The initial conditions are such that the initial position and velocity tracking errors are zero and the vehicle is upside down, that is, the thrust vector is pointing in the direction of the gravity vector. As a result, the vehicle needs to quickly reorient itself in order to track the desired reference trajectory. To achieve this, the initial orientation error  $x_1(0, 0)$  triggers the update of the memory variable  $y_1$  that is represented in Figure 1. It is possible to verify in Figure 4 that the controller switching (which occurs at  $t = 0$ ) leads to an instantaneous decrease of the potential function  $V_1$  and, additionally, the function is strictly decreasing as a function of continuous time, corresponding to a fast attitude correction within the first few seconds of the simulation. While the vehicle reorients itself to track the desired thrust vector, the position and the velocity tracking errors increase but this transient response is soon mitigated and the vehicle converges to the desired trajectory, as can be verified in Figure 3.

#### 4.4 Experimental Results

In this section, we present the experimental results for quadrotor trajectory tracking using the controller designed in Section 4.2 with the synergistic potential function (14). Given  $f > 0$ , the reference trajectory is given

by  $t \mapsto p_d(t) = \begin{bmatrix} \cos 2\pi f t & \sin 2\pi f t & 0 \end{bmatrix}^\top$  for each  $t \geq 0$ .

The reference acceleration and jerk, given by  $\ddot{p}_d$  and  $p_d^{(3)}$ , respectively, satisfy Assumption 1 with  $M_2 = (2\pi f)^2$  and  $M_3 = (2\pi f)^3$ . The thrust vector is  $r_1 = \begin{bmatrix} 0 & 0 & -1 \end{bmatrix}^\top$ ,

the gravity vector is given by  $g = \begin{bmatrix} 0 & 0 & 9.81 \end{bmatrix}^\top$ . Assumption 2 is satisfied with  $\underline{\alpha}$ ,  $\bar{\alpha}$  and  $\lambda_1$  given by (23a), (23b) and (23c), respectively. For  $k = 1$ ,  $\gamma = -0.5$ ,  $\delta = \frac{1+\gamma}{2(\frac{2}{k}+1+\gamma)} = 0.1$  we obtain  $\bar{\alpha} = 0.5$ ,  $\underline{\alpha} = 1/6$ ,  $V^* \leq \frac{2}{2+k(1+\gamma)} + \delta = 0.82$ , and  $\lambda_1 = 0.18$ . We design the position controller following (C1) through (C4) considering  $U := \Omega_H(1)$  with  $\bar{k}_1 = 12$ ,  $k_p = 1$  and

$$\begin{aligned} H &:= \text{diag} \left( \begin{bmatrix} 500 & 500 & 500 & 100 & 100 & 100 \end{bmatrix} \right), \\ \hat{Q} &:= \text{diag} \left( \begin{bmatrix} 10 & 10 & 100 & 100 & 100 & 1 \end{bmatrix} \right), \\ \hat{R} &:= 10I_3. \end{aligned}$$

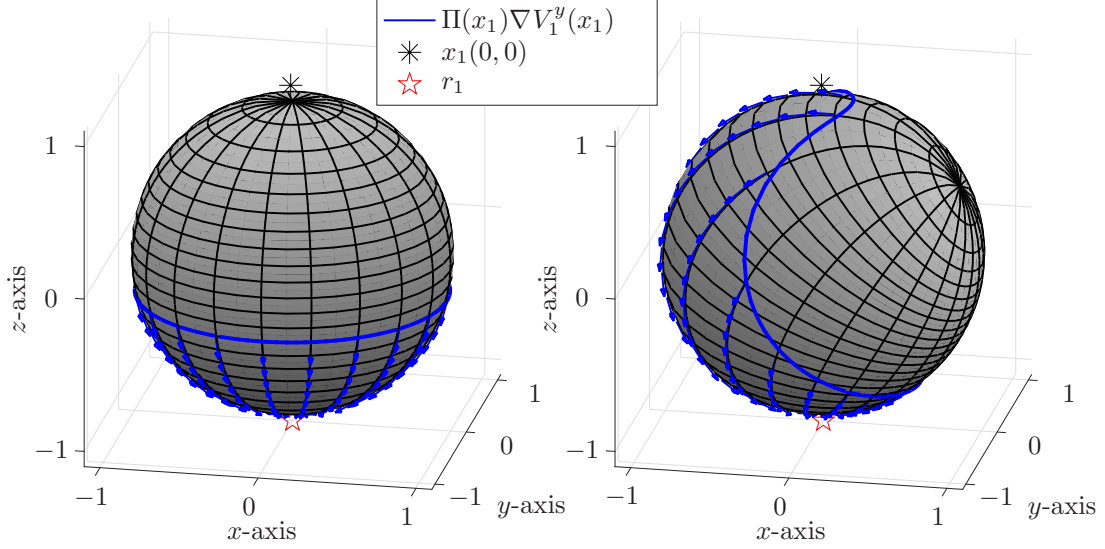


Fig. 1. The arrows represent the gradient vector field of  $x_1 \rightarrow V_1^{y_1}(x_1)$  with  $r_1 = [0 \ 0 \ -1]^T$  and  $x_1 := R\rho_1(z)$  restricted to the set  $\{x_1 \in \mathbb{S}^2 : \mu(x_1, y_1) \leq \delta\}$  for two different values of  $y_1$ : on the left we have  $y_1 = [0 \ 0 \ 1]^T$  and on the right we have  $y_1 = [\sqrt{3}/2 \ 0 \ 0.5]^T$ . The value  $x_1(0,0) \approx [-0.0101 \ 0 \ 0.9999]^T$  lies outside the flow set when the former value of  $y_1$  is used, generating a jump of the logic variable to the latter value.

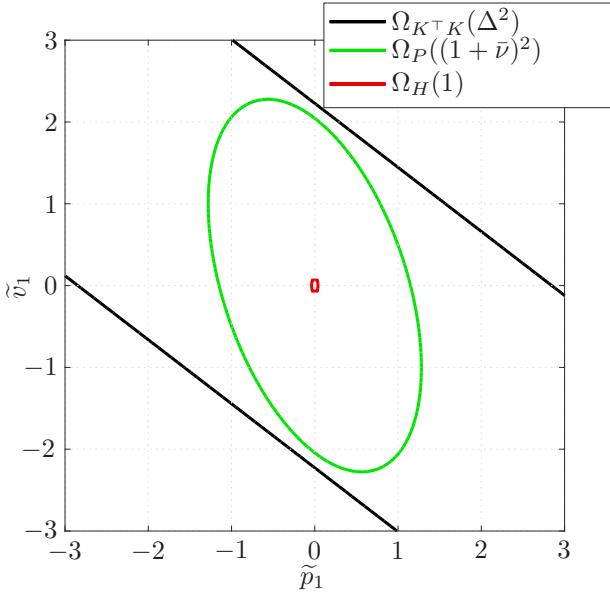


Fig. 2. Representation of a cross-section of the sets  $\Omega_H(1)$ ,  $\Omega_P(\ell_P)$ ,  $\Omega_{K^T K}(\Delta^2)$  and of the solution to the closed-loop hybrid system for the controller parameters and initial conditions presented in Section 4.4. Solutions with initial conditions in  $\Omega_H(1)$  (red inner circle) do not leave  $\Omega_P((1+\bar{\nu})^2)$  (green outer ellipse), hence do not violate the saturation constraints, which would happen if solutions were to cross the dark parallel lines.

The experimental setup is akin to that of [Cabecinhas et al., 2014, Section 6]. To test the proposed controller under the given setup, we conducted a series of experiments which involved throwing the quadrotor up in the air and switching on the controller at the top of its trajectory

using as a reference position that of the moment the controller was switched on. In the experiments, we set  $k_p = 0.05$  in order to compensate for the lag in angular velocity tracking. Alternatively, we could have reduced the controller gains of the position controller  $K$  at the expense of controller performance. Figure 6 shows the results of one of the experimental runs.

The initial position error is zero because the reference position is the quadrotor position when the controller is activated. Note that the initial velocity error is very high because the quadrotor is being thrown into the working area and the reference velocity is set to zero. It is possible to verify that the magnitude of the noise in the velocity measurements is very high in the first few seconds of the experiment. This is due to two main factors: since the quadrotor is being thrown into the working area upside down, there are several reflective markers which disappear from the optical motion capture system field of view, thus reducing the accuracy in the position estimation. These imprecisions are compounded by the fact that the velocity is estimated from the position measurements by determining their rate of change. This noise has an effect in the determination of the desired thrust vector which induces controller switching at points during the first few seconds of the experiment. Nevertheless, the orientation error  $V_1$  has an overall decrease in the first 2.5 s of the experiment and is kept near zero for the larger portion of the experiment which corresponds to zero orientation error. More importantly, the position and velocity converge to zero within the represented time span. This controller can be used together with higher-level controllers to perform aggressive maneuvers as illustrated in Casau et al. [2017a].

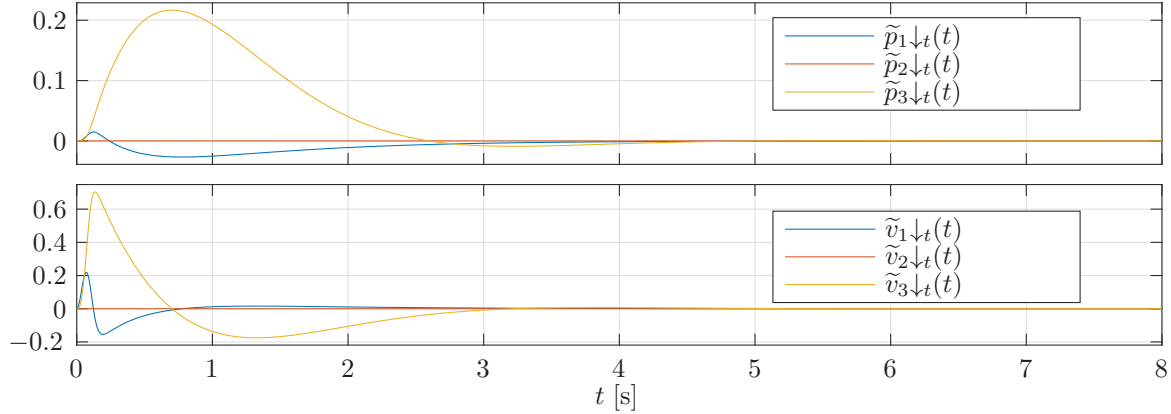


Fig. 3. Simulation results representing the evolution of the position and velocity tracking errors for the closed-loop system with initial conditions and controller parameters given in Section 4.4.

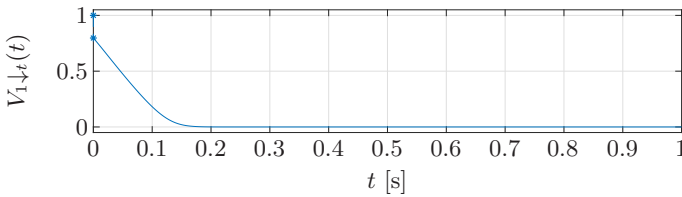


Fig. 4. Simulation results representing the evolution of the potential function  $t \mapsto V_{1\downarrow t}(t) := V_1(R_{\downarrow t}(t)^\top \rho(z_{\downarrow t}(t)), y_{1\downarrow t}(t))$  for the closed-loop system with initial conditions and controller parameters given in Section 4.3.

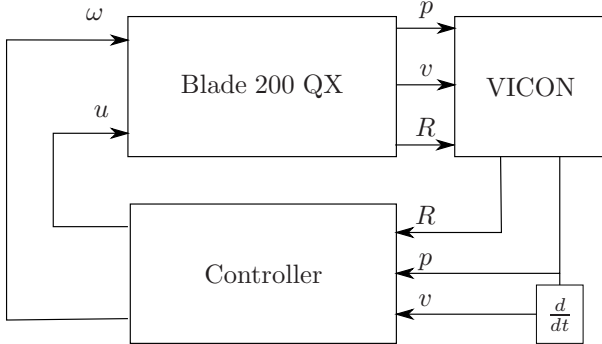


Fig. 5. Diagram representing the experimental setup.

## 5 Conclusions

In this paper, we have demonstrated that the existence of synergistic potential functions on  $\mathbb{S}^n$  is a sufficient condition for the asymptotic stabilization of systems evolving on the  $n$ -dimensional sphere. Moreover, if these functions and their derivatives satisfy some additional bounds, it is possible to achieve global exponential stability. We provided a construction of synergistic potential functions which generates flows along geodesics upon switching. The proposed controller was then applied to trajectory tracking for a vectored thrust vehicle.

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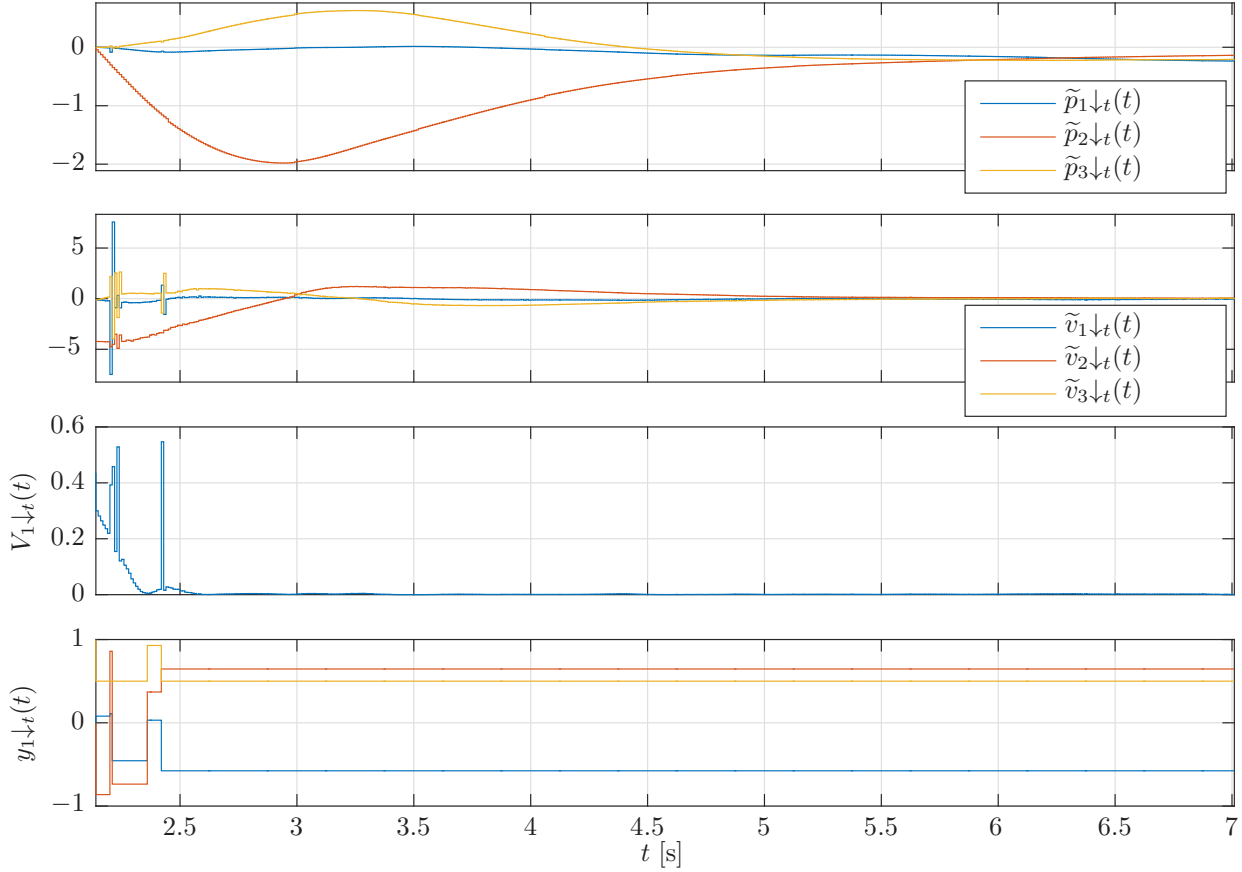


Fig. 6. Experimental results – the quadrotor is thrown into the Vicon workspace and the controller is activated at the top of the trajectory.

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## A Auxiliary results

**Proposition A.1** *Let  $r \in \mathbb{S}^n$  and let  $\mathcal{Q}$  be a compact set. Given  $V \in \mathcal{Q}_r$ , the following holds:*

- (1)  $r \in \text{crit}(V^y)$  for each  $y \in \mathcal{Q}$ ;
- (2) *The function (6) is continuous and the map (10) is outer semicontinuous.*

**PROOF.** Since  $V$  is assumed to be continuously differentiable and  $V^y$  is positive definite relative to  $r$ , this implies that  $r \in \text{crit } V^y$  for each  $y \in \mathcal{Q}$ . From the continuity of  $V$ , we have that (6) is continuous if and only if  $\nu(x) := \min_{\bar{y} \in \mathcal{Q}} V(x, \bar{y})$  is continuous. Let  $\{x_i\}_{i \in \mathbb{N}}$  denote a sequence that converges to some  $x \in \mathbb{S}^n$  and let  $\{y_i\}_{i \in \mathbb{N}}$  be such that  $y_i \in \arg \min_{\bar{y} \in \mathcal{Q}} V(x_i, \bar{y})$ . Since  $\mathcal{Q}$  is compact, there is at least one convergent subsequence of  $\{y_i\}_{i \in \mathbb{N}}$ . For every convergent subsequence, denoted by  $\{y_{k(i)}\}_{i \in \mathbb{N}}$ , with accumulation point  $y_k \in \mathcal{Q}$ , we conclude that  $V(x_{k(i)}, y_{k(i)})$  converges to  $V(x, y_k)$  because  $V$  is continuous. Suppose that there exists  $z \in \mathcal{Q}$  such that  $V(x, z) < V(x, y_k)$  and suppose that  $\{z_i\}_{i \in \mathbb{N}}$  is a sequence that converges to  $z$ . By the continuity of  $V$  and by the assumption that  $V(x, z) < V(x, y_k)$ , there exists  $I \in \mathbb{N}$  such that  $V(x_{k(i)}, z_{k(i)}) < V(x_{k(i)}, y_{k(i)})$  for each  $i \geq I$ . However, this is a contradiction because  $y_{k(i)}$  is the minimizer for  $V(x_{k(i)}, y_{k(i)})$ . We conclude that  $\lim_{i \rightarrow \infty} \nu(x_i) = \nu(x)$ , thus  $\nu(x)$  is continuous. Also, for every sequence  $\{x_i\}_{i \in \mathbb{N}}$  convergent to  $x \in \mathbb{S}^n$ , and for each sequence  $\{y_i\}_{i \in \mathbb{N}}$  convergent to  $y \in \mathcal{Q}$  such that  $y_i \in \rho(x_i)$  for each  $i \in \mathbb{N}$ , we have that  $y \in \rho(x)$  which proves that  $\rho(x)$  is outer semicontinuous.  $\square$

**Lemma A.1** *The function  $V : \mathcal{V} \rightarrow \mathbb{R}$  satisfies*

$$\arg \min_{(x,y) \in \mathcal{V}} V(x,y) = V^{-1}(0) = \{(r,y) \in \mathcal{V}\} \quad (\text{A.1})$$

$$\arg \max_{(x,y) \in \mathcal{V}} V(x,y) = V^{-1}(1) = \{(x,x) \in \mathcal{V}\} \quad (\text{A.2})$$

*Moreover,  $V(x,y)$  is positive definite on  $\mathcal{V}$  relative to  $\{(r,y) \in \mathcal{V}\}$  and for each  $x \in \mathbb{S}^n$ ,*

$$\arg \min_{y \in \mathbb{S}^n} V(x,y) = \begin{cases} -x & \text{if } x \neq r \\ \mathbb{S}^n \setminus \{r\} & \text{if } x = r \end{cases} \quad (\text{A.3})$$

$$\min_{y \in \mathbb{S}^n} V(x,y) = \frac{1 - r^\top x}{2k + 1 - r^\top x}. \quad (\text{A.4})$$



**PROOF.** Since  $h_r(x) \geq 0$  for all  $x \in \mathbb{S}^n$  and  $h_r(x) = 0$  if and only if  $x = r$ , it follows that  $h_r(x) + kh_y(x) > 0$  on  $\mathcal{V}$ . Setting  $V(x, y) \leq 0$ , we find that  $h_r(x) \leq 0$ , which can only be satisfied (with equality) when  $x = r$ . That is,  $V$  attains its minimum value of zero on the set  $\{(r, y) \in \mathcal{V}\}$ . Similarly,  $V(x, y) \geq 1$  if and only if  $h_y(x) \leq 0$ , which is again satisfied (with equality) only when  $x = y$  and thus,  $V$  attains its maximum value of one on the set  $\{(x, x) \in \mathcal{V}\}$ . To prove (A.3) and (A.4), we note—from our previous observations—that

$$\min_{\bar{y} \in \mathbb{S}^n} V(x, \bar{y}) = \frac{1 - r^\top x}{1 - r^\top x + k \max_{\bar{y} \in \mathbb{S}^n} (1 - \bar{y}^\top x)}. \quad (\text{A.5})$$

Since, for each  $x \in \mathbb{S}^n$ ,  $1 - r^\top x$  attains its maximal value of two at  $y = -x$ , this choice minimizes  $y \mapsto V(x, y)$  for each  $x \in \mathbb{S}^n \setminus \{r\}$ ; however, when  $x = r$ ,  $V(x, y) = 0$  for any  $y \in \mathbb{S}^n \setminus \{r\}$ , thus proving (A.3). Eq. (A.4) follows from evaluating  $V(x, y)$  with  $y = -x$ .  $\square$

**Corollary A.1** *Given  $y \in \mathbb{S}^n$ , define  $V^y : \mathbb{S}^n \rightarrow \mathbb{R}$  for each  $x \in \mathbb{S}^n$  as  $V^y(x) = V(x, y)$ , with  $V$  given by (14). Then,*

$$\text{crit } V^y = \begin{cases} \{r, y\} & \text{if } r \neq y \\ \mathbb{S}^n & \text{otherwise.} \end{cases}$$

**PROOF.** By definition,  $x \in \text{crit } V^y$  if and only if  $\Pi(x) \nabla V^y(x) = 0$ , or equivalently,  $|\Pi(x) \nabla V^y(x)| = 0$ . Noting that  $V^y(x) = V(x, y)$  for each  $x \in \mathbb{S}^n$ , it follows that  $\nabla V^y(x) = \nabla_r V(x, y)$  for each  $x \in \mathbb{S}^n$ . By (15b) of Lemma 1, it follows that  $x \in \text{crit } V^y$  if and only if

$$V^y(x)(1 - V^y(x))(1 - r^\top y) = 0. \quad (\text{A.6})$$

Clearly, (A.6) is satisfied in three cases:  $V^y(x) = 0$ ,  $1 - V^y(x) = 0$ , or  $1 - r^\top y = 0$ . When  $1 - r^\top y = 0$ , or equivalently, when  $r = y$ , it follows that every point in  $\mathbb{S}^n$  is a critical point, since (A.6) is satisfied for every  $x \in \mathbb{S}^n$ . In fact, when  $r = y$ ,  $V^y(x) = 1/(1 + k)$ , so that for all  $x \in \mathbb{S}^n$ ,  $\nabla V^y(x) = 0$  and obviously  $\Pi(x) \nabla V^y(x) = 0$ . We now examine the remaining cases. When  $V^y(x) = 0$ , it follows from the definition of  $V^y(x) = V(x, y)$  that  $1 - r^\top x = 0$ , or equivalently,  $x = r$ . If  $V^y(x) = 1$ , a short calculation yields  $1 - y^\top x = 0$ , or equivalently,  $x = y$ . Thus, when  $r \neq y$ ,  $\text{crit } V^y = \{r, y\}$ .  $\square$

**Lemma A.2** *The following holds*

$$0 < 1 + k - \sqrt{1 + 2k\gamma + k^2} \leq 1 - r^\top x + k(1 - y^\top x) \leq 1 + k + \sqrt{1 + 2k\gamma + k^2}. \quad (\text{A.7})$$

for all  $(x, y) \in \mathbb{S}^n \times \mathcal{Q}$ .

**PROOF.** The upper and lower bounds on (A.7) follow from the solution to the optimization problem  $\min / \max \{J(x, y) : 1 - y^\top x = 0, 1 - x^\top x = 0, r^\top y - \gamma =$

$0\}$  with  $J(x, y) = 1 - r^\top x + k(1 - y^\top x)$  for each  $(x, y) \in \mathbb{S}^n \times \mathcal{Q}$ , by means of Lagrange multipliers. To prove this bound, we solve the following optimization problem:

$$\begin{aligned} \min / \max \quad & J(x, y) \\ \text{subject to} \quad & 1 - y^\top y = 0 \\ & 1 - x^\top x = 0 \\ & r^\top y - \gamma = 0, \end{aligned} \quad (\text{A.8})$$

where  $J(x, y) = 1 - r^\top x + k(1 - y^\top x)$  for each  $(x, y) \in \mathbb{S}^n \times \mathcal{Q}$ . We proceed by constructing the Lagrangian,  $L : \mathbb{S}^n \times \mathbb{Q} \times \mathbb{R} \times \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$  for each  $(x, y, \lambda_1, \lambda_2, \beta) \in \mathbb{S}^n \times \mathcal{Q} \times \mathbb{R} \times \mathbb{R} \times \mathbb{R}$  as

$$\begin{aligned} L(x, y, \lambda_1, \lambda_2, \beta) = & J(x, y) + \frac{\lambda_1}{2}(1 - y^\top y) \\ & + \frac{\lambda_2}{2}(1 - x^\top x) + \beta(r^\top y - \gamma). \end{aligned}$$

According to the KKT conditions, each solution to (A.8) must satisfy  $\nabla L = 0$ ,  $\beta \geq 0$ , equivalently

$$\lambda_1 y = \beta r - kx \quad (\text{A.9})$$

$$\lambda_2 x = -r - ky, \quad (\text{A.10})$$

in addition to  $\beta \geq 0$  and

$$\beta(r^\top y - \gamma) = 0. \quad (\text{A.11})$$

We break the analysis into cases. Assuming that  $\beta = 0$ , it follows from (A.9) and (A.10) that  $x = -y = r$ , where  $J(r, -r) = 2k$  or  $x = y = -r$ , where  $J(-r, -r) = 2$ .

Assuming that  $\beta \neq 0$ , it follows from (A.11) that  $r^\top y = \gamma$ . Then, by multiplying (A.10) on the left by  $x^\top$ , it follows that

$$\lambda_2 = -r^\top x - ky^\top x$$

so that  $J(x, y) = 1 + k + \lambda_2$ . In the same direction, by taking the norm of both sides of (A.9) and applying the fact that  $r^\top y = \gamma$ , it follows that

$$\lambda_2^2 = 1 + 2k\gamma + k^2,$$

so that  $J(x, y) = 1 + k \pm \sqrt{1 + 2k\gamma + k^2}$ .

Now that we have all possible values of  $J(x, y)$  that solve (A.8), we evaluate to find the minimum and maximum of  $J$ . Clearly,  $1 + k - \sqrt{1 + 2k\gamma + k^2} < 1 + k + \sqrt{1 + 2k\gamma + k^2}$ . Moreover, it is obvious that  $1 \leq \sqrt{1 + 2k\gamma + k^2}$  and  $k \leq \sqrt{1 + 2k\gamma + k^2}$ , thus  $\max\{2, 2k\} \leq 1 + k + \sqrt{1 + 2k\gamma + k^2}$ .  $\square$